

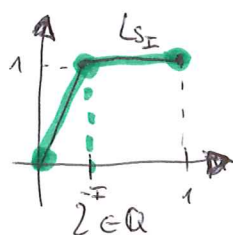
Single-shot Thermo, lecture 6

25.11.
2014

- There are now handwritten lecture notes online.
→ correction to lecture 2!
- Exercises 5: now 4 problems, but 2 weeks of time to solve them!
- Quick repetition from last time:

Sharp state S_I : if $I \in \mathbb{N}$, then $S_I = \begin{pmatrix} 1 \\ 0 \end{pmatrix}^{\otimes I}$ ("I pure bits");

otherwise



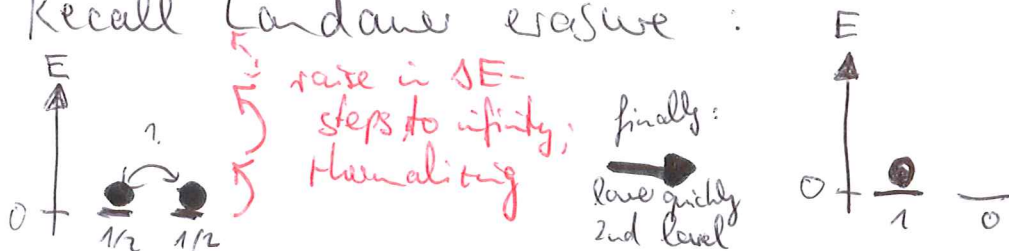
Distillable nonuniformity: $p \xrightarrow{\text{noisy}} S_I$ iff $I \leq I_0(p) := \log d_p - H_0(p)$

Nonuniformity of formation:

$S_I \xrightarrow{\text{noisy}} p$ iff $I \geq I_\infty(p) := \log d_p - H_\infty(p)$.

2.8. Interpreting I_∞ (and I_0): a clarification

Recall Landauer erasure:



Expected work $\langle W \rangle = k_B T \log 2$

Shown by Browne, Garner, Dahlste, Vedral; arXiv: 1311.7612:

Prob(W)



width of distribution $\sigma \leq \frac{1}{2} \Delta E \sqrt{N}$

N : # of raising steps;

● $\Delta E \rightarrow 0$ faster than $N \rightarrow \infty$ can have $\sigma \rightarrow 0$.

①

\Rightarrow investing $k_B T \log 2 + \delta$ of work, with $\delta > 0$ arbitrarily small, we can make almost 100% sure to reset the bit.

Homework exercise 5: Suppose we don't want $\begin{pmatrix} 1/2 \\ 1/2 \end{pmatrix} \rightarrow \begin{pmatrix} 1 \\ 0 \end{pmatrix}$ but $\begin{pmatrix} 1/2 \\ 1/2 \end{pmatrix} \rightarrow \begin{pmatrix} 1-p \\ p \end{pmatrix}$, $0 \leq p \leq \frac{1}{2}$. (don't raise ϵ all up to infinity!)

Then $\langle W \rangle = k_B T \cdot \underbrace{I_\infty(p)}_{\text{"work of formation of } p} = k_B T [\log 2 - H_\infty(p)]$

and work distribution is still arbitrarily peaked!

So what about standard Shannon/von Neuman entropy?

If we have $n \in \mathbb{N}$ (n large) independent particles,

distributed as $\begin{pmatrix} 1/2 \\ 1/2 \end{pmatrix}^{\otimes n}$
 \uparrow
 $\begin{pmatrix} 1/2 \\ 1/2 \end{pmatrix} \dots \begin{pmatrix} 1/2 \\ 1/2 \end{pmatrix}$ and we want to get them distributed close to $\begin{pmatrix} 1-p \\ p \end{pmatrix}^{\otimes n}$

Then $\langle W \rangle = k_B T (n \cdot H(p) + o(n))$

(we will prove this soon).

Similar reasoning applies to Szilard engines and distillable work $I_0(p)$.

2.9. Approximate formation and distillation

Write $p \xrightarrow{\epsilon\text{-noisy}} q$ (where $\epsilon > 0$) if there exists a prob. vector \tilde{q} with $D(q, \tilde{q}) \leq \epsilon$ and $p \xrightarrow{\text{noisy}} \tilde{q}$.

(2) \leftarrow trace distance / variation distance

Consistency of quantum and classical cases:

Lemma: For quantum states ρ, σ , there exists a noisy (quantum) operation \mathcal{N} and another state $\tilde{\sigma}$ with $D(\sigma, \tilde{\sigma}) \leq \varepsilon$ and $\mathcal{N}(\rho) = \tilde{\sigma}$ if and only if $\lambda(\rho) \xrightarrow{\varepsilon\text{-noisy}} \lambda(\sigma)$.

Proof not here.

Definition ("smooth entropies")

For a prob. vector p and $\varepsilon > 0$, define

$$H_{\infty}^{\varepsilon}(p) := \max_{p': D(p, p') \leq \varepsilon} H_{\infty}(p'), \text{ and}$$

$$H_0^{\varepsilon}(p) := \min_{p': D(p, p') \leq \varepsilon} H_0(p').$$

Smooth entropies (conditional versions etc.) have many applications in classical and quantum information theory.

Nonuniformity of approximate formation:

Lemma: The minimum I with $S_I \xrightarrow{\varepsilon\text{-noisy}} p$ is given by

$$I_{\infty}^{\varepsilon}(p) := \log d_p - H_{\infty}^{\varepsilon}(p).$$

Proof: Suppose $S_I \xrightarrow{\text{noisy}} p'$. Minimal I needed for this is

$$I_{\infty}(p') = \log \underbrace{d_p}_{=d_{p'}} - H_{\infty}(p').$$

Minimize this over all p' with $D(p, p') \leq \varepsilon$:

$$\min_{p': D(p, p') \leq \varepsilon} I_{\infty}(p') = \log d_p - \max_{p': D(p, p') \leq \varepsilon} H_{\infty}(p')$$

$$= \log d_p - H_{\infty}^{\varepsilon}(p). \quad \square$$

Interpretation: $k_B T I_{\infty}^{\varepsilon}(p)$ is the amount of work we have to invest to create p approximately / with failure probability $\leq \varepsilon$. (cf. explanations in 2.7 / interpretation of trace distance.)

Approximate distillable nonuniformity:

Lemma: The maximum I with $p \xrightarrow{\varepsilon\text{-noisy}} S_{\mathbb{I}}$ is at least

$$I_0^{\varepsilon}(p) := \log d_p - H_0^{\varepsilon}(p).$$

(Exact value: later.)

Proof: Choose p' with $D(p, p') \leq \varepsilon$ and $H_0^{\varepsilon}(p) = H_0(p')$.

Then there exists a noisy operation \mathcal{N} such that

$$\mathcal{N}(p') = S_{\mathbb{I}}, \text{ where } \mathbb{I} = I_0(p') = I_0^{\varepsilon}(p).$$

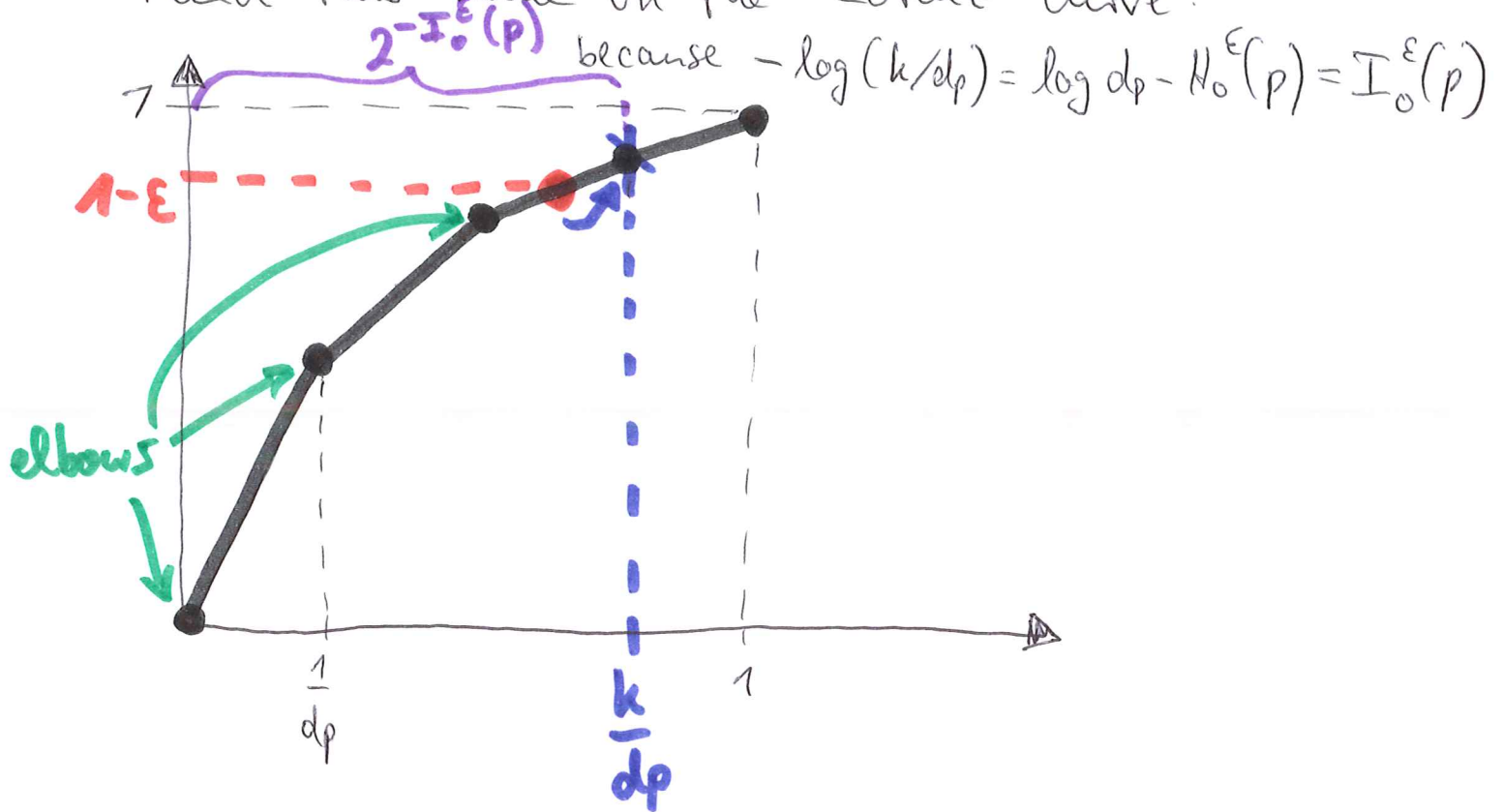
Apply \mathcal{N} to p instead, and set $q := \mathcal{N}(p)$, we get

$$D(q, S_{\mathbb{I}}) = D(\mathcal{N}(p), \mathcal{N}(p')) \leq D(p, p') \leq \varepsilon.$$

Thus $p \xrightarrow{\varepsilon\text{-noisy}} S_{\mathbb{I}}$ for $\mathbb{I} = I_0^{\varepsilon}(p)$. \square

Homework: $H_0^{\varepsilon}(p) = \log k$, where $k \geq 1$ is the smallest integer with $\sum_{i=1}^k x_i^{\downarrow} \geq 1 - \varepsilon$.

Find this value on the Lorenz curve:



Lemma: The maximum I with $p \xrightarrow{\epsilon\text{-noisy}} S_I$ is exactly

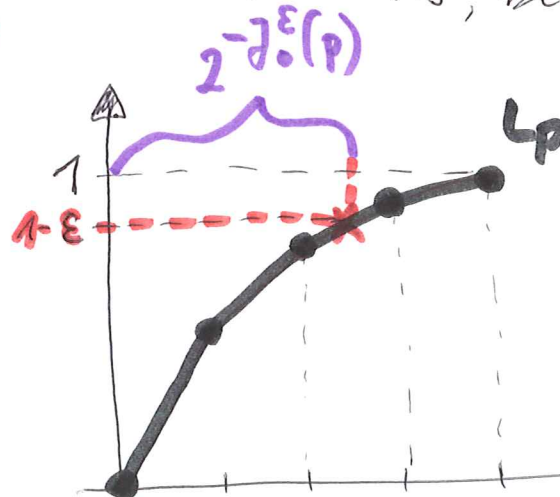
$$J_0^\epsilon(p) := \lim_{N \rightarrow \infty} I_0^\epsilon(p \otimes \gamma_N)$$

\uparrow
 $(\frac{1}{N_1} \rightarrow \frac{1}{N})$

" \geq " is clear because p and $p \otimes \gamma_N$ are "noisy-equivalent";
 " \leq " needs more work to prove (not here).

p and $p \otimes \gamma_N$: same Lorenz curves, but $p \otimes \gamma_N$ has "more elbows"

$\Rightarrow J_0^\epsilon$ easy to read off:



Discuss why.

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2.10. State conversion in the thermodynamic limit, recovering Shannon and von Neuman entropy

Thermodynamic limit: we have n systems/particles that are not strongly correlated, and n is large. Ideally, we have $p^{\otimes n}$, and send $n \rightarrow \infty$.

Lemma (asymptotic equipartition property):

For every $0 < \epsilon < 1$, we have

$$\lim_{n \rightarrow \infty} \frac{1}{n} H_0^\epsilon(p^{\otimes n}) = H(p), \text{ and}$$

$$\lim_{n \rightarrow \infty} \frac{1}{n} H_\infty^\epsilon(p^{\otimes n}) = H(p),$$

where H is Shannon entropy, $H(p) = -\sum_i p_i \log p_i$.

Proof: Next time. Follows from Shannon's noiseless coding theorem (data compression!).

Homework: you show the following:

leave on blackboard.

Lemma: If $I_0^{\epsilon/2}(p) \geq I_\infty^{\epsilon/2}(q)$ then $p \xrightarrow{\epsilon\text{-loisly}} q$.

We will use this to answer the following

Questions: Suppose we have many copies of p . How many copies of q can we distill?

• Suppose we want many copies of q . How many copies of p do we need?

Theorem: p, q prob. vectors, not maximally mixed, $0 < \epsilon < 1$. For every $n \in \mathbb{N}$, let m_n be the largest integer such that

$$p^{\otimes n} \xrightarrow{\epsilon\text{-noisy}} q^{\otimes m_n}$$

Then
$$\lim_{n \rightarrow \infty} \frac{m_n}{n} = \frac{I(p)}{I(q)}$$

with $I(p) := \log d_p - H(p)$ the (Shannon) negentropy. Similarly, let k_n be the smallest integer such that

$$p^{\otimes k_n} \xrightarrow{\epsilon\text{-noisy}} q^{\otimes n}$$

Then
$$\lim_{n \rightarrow \infty} \frac{k_n}{n} = \frac{I(q)}{I(p)}$$

Interpretation:

- In the thermodynamic limit, negentropy I is the "unique measure of information".

- "Distillable nonuniformity": $p^{\otimes n} \xrightarrow{\epsilon\text{-noisy}} \underbrace{\begin{pmatrix} 1 \\ 0 \end{pmatrix}}_{\text{pure}}^{\otimes m_n}$

works for $m_n = n \cdot I(p) + o(n)$

\Rightarrow get $\approx n \cdot I(p)$ pure bits from $p^{\otimes n}$.

- "Nonuniformity of formation": $\begin{pmatrix} 1 \\ 0 \end{pmatrix}^{\otimes k_n} \xrightarrow{\epsilon\text{-noisy}} p^{\otimes n}$

works for $k_n = n \cdot I(p) + o(n)$

\Rightarrow need $\approx n \cdot I(p)$ pure bits to create $p^{\otimes n}$.

\Rightarrow Noisy operators become "reversible" in the thermodynamic limit!

However, in the microscopic / single-shot regime, we have irreversibility due to $\underbrace{I_0^\varepsilon(p)}_{\text{distillable}} \ll \underbrace{I_\infty^\varepsilon(p)}_{\text{formation}}$ in general.

Proof of theorem: Lemma on BB \Rightarrow

$$\text{if } \frac{m}{n} \leq \frac{\frac{1}{n} I_0^{\varepsilon/2}(p^{\otimes n})}{\frac{1}{m} I_\infty^{\varepsilon/2}(q^{\otimes m})} \text{ then } p^{\otimes n} \xrightarrow{\varepsilon\text{-noisy}} q^{\otimes m}$$

But this is impossible for $m = m_n + 1$

$$\Rightarrow \frac{m_n + 1}{n} > \frac{\frac{1}{n} I_0^{\varepsilon/2}(p^{\otimes n})}{\frac{1}{m_n + 1} I_\infty^{\varepsilon/2}(q^{\otimes (m_n + 1)})}$$

Easy to see: $\lim_{n \rightarrow \infty} m_n = \infty$, so

$$\begin{aligned} \liminf_{n \rightarrow \infty} \frac{m_n}{n} &\geq \liminf_{n \rightarrow \infty} \frac{\frac{1}{n} I_0^{\varepsilon/2}(p^{\otimes n})}{\frac{1}{m_n + 1} I_\infty^{\varepsilon/2}(q^{\otimes (m_n + 1)})} \\ &= \frac{\log d_p - \lim_{n \rightarrow \infty} \frac{1}{n} H_0^{\varepsilon/2}(p^{\otimes n})}{\log d_q - \lim_{n \rightarrow \infty} \frac{1}{m} H_\infty^{\varepsilon/2}(q^{\otimes m})} = \frac{I(p)}{I(q)}. \end{aligned}$$

We have $p^{\otimes n} \xrightarrow{\text{noisy}} q_n$ with $D(q^{\otimes m_n}, q_n) \leq \varepsilon$.

Choose $\delta > 0$ such that $\varepsilon + \delta < 1$.

8 Homework: I_∞^δ is a nonuniformity monotone \Rightarrow

$$I_{\infty}^{\delta}(p^{\otimes n}) \geq I_{\infty}^{\delta}(q_n) = I_{\infty}(q'_n)$$

$$\text{with } D(q_n, q'_n) \leq \delta.$$

$$\Rightarrow D(q^{\otimes m_n}, q'_n) \leq D(q^{\otimes m_n}, q_n) + D(q_n, q'_n) \leq \varepsilon + \delta$$

$$\Rightarrow I_{\infty}^{\delta+\varepsilon}(q^{\otimes m_n}) = \min_{q': D(q^{\otimes m_n}, q') \leq \delta+\varepsilon} I_{\infty}(q')$$

$$\leq I_{\infty}(q'_n) = I_{\infty}^{\delta}(q_n) \leq I_{\infty}^{\delta}(p^{\otimes n}).$$

$$\Rightarrow \frac{m_n}{n} \leq \frac{\frac{1}{n} I_{\infty}^{\delta}(p^{\otimes n})}{\frac{1}{m_n} I_{\infty}^{\delta+\varepsilon}(q^{\otimes m_n})}$$

$$\Rightarrow \limsup_{n \rightarrow \infty} \frac{m_n}{n} \leq \frac{\lim_{n \rightarrow \infty} \frac{1}{n} I_{\infty}^{\delta}(p^{\otimes n})}{\lim_{n \rightarrow \infty} \frac{1}{m_n} I_{\infty}^{\delta+\varepsilon}(q^{\otimes m_n})} = \frac{I(p)}{I(q)}.$$

Second statement can be proven analogously. \square