

Quantum Information Theory

What is the information content of a state?

- Information is "what we can learn" (Claude Shannon)

• Suppose we flip a coin. What do we learn about odds of H vs T?

+ If we flip a fair coin, we don't get a lot of information for each H or T. We know $P(H) \approx P(T) \approx \frac{1}{2}$ reasonably quickly, but it takes a while to nail down.

+ If $P(H) \gg P(T)$, each H tells us nearly nothing, but each T tells us a lot about $P(T)$.

+ Information is about classical ignorance

• Quantitatively, suppose we flip N coins. We expect $P(H)N = pN$ heads and $(1-p)N$ tails.

+ The number of arrangements is (Stirling's Approx)

$$\# = \frac{N!}{(pN)!(1-p)N!} \approx \frac{N^N}{(pN)^{pN}(1-p)N^{(1-p)N}} = \frac{1}{p^{pN}(1-p)^{(1-p)N}}$$

+ We define Shannon entropy S such that $\# = 2^S$
so

$$S = -p \log_2 p - (1-p) \log_2 (1-p)$$

This is like "how much" of a coin each flip contributes ← There are 2^S alternatives per flip

+ If there are multiple options w/probabilities P_i each, then

$$S = -\sum P_i \log_2 P_i$$

+ The entropy is the average of the information content of each flip value $i = -\log_2 P$ in Shannon bits.

- Alternatively, we can think in terms of ^{data} compression
 - + Consider an alphabet w/ 4 letters a, b, c, d
 - + To encode these in binary, we can use
 $a = 00, b = 01, c = 10, d = 11$
 - + However, this is inefficient if some letters are more common in the language.
 - + Suppose the probabilities for a given letter to take each value is
 $P(a) = \frac{1}{2}, P(b) = \frac{1}{4}, P(c) = P(d) = \frac{1}{8}$
 - The (most) efficient encoding is
 $a = 0, b = 10, c = 110, d = 111$
 - with the rule that 0 means "next letter" and max length = 3
 - + Then the length of each letter is $i = -\log_2 P$
 - and the Shannon entropy is the average letter length in the language

• Notes:

- + The definition for Shannon entropy is very similar to the Gibbs definition of the thermodynamic entropy (in terms of microstate probability)
- + There are many extensions in classical information theory, but we'll focus on the quantum theory.

- Quantum Entropy

- Quantum information is also what we can learn
 - + If we have a pure state (and know what it is), you know all there is to know about the system even if you can't predict a specific measurement

- + In a mixed state, we can generalize the Shannon entropy to

$$S = -\sum_i P_i \log_2 P_i$$

where the P_i are the eigenvalues of the density matrix. This is the

+ We can write this in terms of the density matrix

$$S = -\sum_i \langle e_i | \rho \log_2(\rho) | e_i \rangle = -\text{tr}(\rho \log_2(\rho))$$

where the sum is over the e^i basis of ρ .

This is the von Neumann entropy

+ Von Neumann entropy of a pure state is $S=0$ ($P_i=1, P_{ij}=0$)

When the mixed state arises as part of an entangled state, von Neumann entropy may be called entanglement entropy

- Properties of von Neumann entropy (mostly presented w/o proof)

- + If the dimension of Hilbert space is N , entropy has max value $S \leq \log_2(N)$, with equality only when the state is maximally mixed — all N eigenvalues of ρ are $P=1/N$.

+ Suppose system 1 is in a mixed state, and we purify it by entangling it with system 2 (systems 1+2 together are in a pure state). Then $S(\rho)=S(\rho_2)$ for the entropies if we partial trace either system.

This is true even if systems 1+2 have different Hilbert space dimensions.

+ The entropy is concave (2nd derivative negative for normal function)

$$\sum_i x_i S(\rho_i) \leq S(\sum_i x_i \rho_i) \text{ where } x_i \geq 0 \text{ and } \sum_i x_i = 1$$

That is, mixing density matrices increases entropy.
(See reading for proofs)

- + Suppose you have a quantum system with 2 pieces.

If the overall density matrix is ρ ,

$$S(\rho) \geq |S(\rho_1) - S(\rho_2)| \quad \text{Araki-Lieb inequality}$$

and

$$S(\rho) \leq S(\rho_1) + S(\rho_2) \quad \text{subadditivity of entropy}$$

where $\rho_1 + \rho_2$ are the reduced density operators

+ If a system has 3 parts 1, 2, + 3, Then

$$S(\rho_{1,2}) + S(\rho_{2,3}) \geq S(\rho_1) + S(\rho_2)$$

where ρ = overall density matrix

$\rho_{1,2}$ = reduced by partial trace over 3

$\rho_{2,3}$ = reduced by partial trace over 1

ρ_2 = reduced by partial trace over 1 + 3

This is strong subadditivity

• Other "entropies"

+ Many of these properties are proven using the relative entropy

$$S(\rho \| \sigma) = \text{tr} [\rho (\log_2 \rho - \log_2 \sigma)]$$

where ρ = density operator of the system and

σ is an arbitrary reference density operator.

The key property is that $S(\rho \| \sigma) \geq 0$.

+ The Rényi entropy is defined for a number α as

$$S_\alpha(\rho) = \frac{1}{1-\alpha} \log_2 (\text{tr} \rho^\alpha)$$

This becomes von Neumann entropy in the limit $\alpha \rightarrow 1$

+ This is only a very short introduction and there is a lot more you can learn!