

Recall: $H(X)$ measures the ignorance on the rv X .

Let X, Y be two rv's, with distribution $p(xy)$.
 $H(XY) = H(p)$ as before (treat XY as a composite rv).

Fix a particular outcome for Y , say y , with X unknown.

Define $q_y = p(X|Y=y)$ as the distribution of X given $Y=y$.

$$q_y(x) = p(xy) / \sum_x p(xy)$$

$H(q_y)$ is the uncertainty of X when $Y=y$.

Def: Conditional entropy $H(X|Y) = \sum_y p(y) H(q_y)$.

Meaning: average (over unknown Y) uncertainty of X :

easy to remember consequence \rightarrow Fact: $H(X|Y) = H(XY) - H(Y)$
 aka "Chain rule." Proof: def+algebra

Fact: $H(X|Y) = H(XY) - H(Y)$.

Average (over y) uncertainty of X given Y .

Def [mutual information]: $I(X:Y) = H(X) - H(X|Y)$

uncertainty of X before after
 conditioning on Y

i.e. it equals to the information about X contained in Y
 = decrease in uncertainty of X due to conditioning on Y .

By "fact": $I(X:Y) = H(X) + H(Y) - H(XY) = I(Y:X)$

$I(X:Y)$ is MUTUAL (information) between X & Y .

Properties of $H(X)$, $H(X|Y)$, $I(X:Y)$:

- $0 \leq H(X) \leq \log |\Omega|$ [obvious, but useful]
- $H(XY) \leq H(X) + H(Y)$ [called Subadditivity]
 equivalent to $I(X:Y) \geq 0$
 equivalent to $H(X|Y) \leq H(X)$
 [meaning: conditioning reduces uncertainty
 knowing Y cannot hurt]
 "=" iff X, Y independent ($MI=0$, conditioning useless)

Proof of SA and equality condition: p505 Nielsen & Chuang.

Ideas: (i) define relative entropy $H(p||q) = \sum_x p(x) \log [p(x)/q(x)]$.
 (ii) show that it is nonnegative [since $-(\ln 2) (\log z) \geq 1-z$,
 $H(p||q) = -\sum_x p(x) \log[q(x)/p(x)] \geq \sum_x p(x) [1-q(x)/p(x)]/(\ln 2) = 0$,
 with equality hold only iff $q(x)=p(x) \forall x$]. (iii) rewriting $I(X:Y)$ as
 $H(p(x,y)||p(x)q(y))$.

Properties of $H(X)$, $H(X|Y)$, $I(X:Y)$:

- Let X_1, X_2 , be (different) rv's with the same Ω
 $H(\sum_k p(k) X_k) \geq \sum_k p_k H(X_k)$

X = rv obtained by average entropy of X_k
 (1) draw k , (2) draw from X_k

[entropy of the average \geq average entropy]
 [follows from (2): LHS = $H(X)$, RHS = $H(X|K)$]

Properties of $H(X)$, $H(X|Y)$, $I(X:Y)$:

- $H(X|Y) \geq 0$ [follows from Def: average over nonnegative entropies]
- $H(XY) = H(Y) + H(X|Y)$ [Chain Rule, extends to multiple rv's]
- $H(XY) \geq H(Y)$ [follows from 4&5]
- $H(Z) + H(XYZ) \leq H(XZ) + H(YZ)$
 [called Strong Subadditivity SSA]
 Note that Z special, XY symmetric.
 As if Z added to each term in SA. (Thus the name)
 equiv to $H(Y|ZX) \leq H(Y|Z)$ or $H(X|ZY) \leq H(X|Z)$
 Conditioning (on a new rv) decreases conditional entropy.
 Here: $H(Y|Z)$ on X or $H(Y|Z)$ on Y .

Quantum analogues:

Recall $S(\rho) = H(\text{spec}(\rho))$

Let A, B be two quantum systems
 ρ density matrix representing state on AB

$$S(AB) = S(\rho), S(A) = S(\text{tr}_B \rho), S(B) = S(\text{tr}_A \rho).$$

Classical: $H(X|Y) = H(XY) - H(Y)$

In quantum setting, no obvious meaning to condition on one of the two systems.

Def: $S(A|B) = S(AB) - S(B)$ Imitate classical expression but not the meaning.

Quantum analogues:

Classical: $I(X:Y) = H(X) - H(X|Y)$

Def [quantum mutual information]:

$$S(A:B) = S(A) - S(A|B) = S(A) + S(B) - S(AB).$$

Imitate classical expression, due to $S(A|B)$, meaning of $S(A:B)$ not immediately clear. (Investigate later.)

Example 1:

Suppose we have a pure state $|\psi\rangle$ on AB.

There is always a Schmidt decomposition:

$$|\psi\rangle = \sum_x \sqrt{p(x)} |e_x\rangle_A |f_x\rangle_B$$

where $\{|e_x\rangle\}$ is orthonormal in C^A , $\{|f_x\rangle\}$ o.n. in C^B .

$$\text{Note that } \rho_A = \sum_x p(x) |e_x\rangle\langle e_x|, \quad \rho_B = \sum_x p(x) |f_x\rangle\langle f_x|$$

$$S(AB) = 0.$$

$$S(A) = S(B) = H(p).$$

$$S(A:B) = 2 H(p), \quad S(A|B) = -H(p)$$

Example 2: Consider a density matrix $\rho = \sum_x \lambda_x |e_x\rangle\langle e_x|$

Suppose we measure in some basis (WLOG the computation basis) and the outcome is y.

$$\text{Let } |e_x\rangle = \sum_y V_{xy} |y\rangle.$$

Make a matrix V where V_{xy} = entry for the x-col & y-row so V transforms the comp basis to the eigenbasis of ρ .

$$\text{Note that } p(y) = \sum_x \lambda_x |V_{xy}|^2 \text{ and } p(y|x) = |V_{xy}|^2$$

The distributed given by $p(y)$ (as a vector labeled by y) is obtained from the distribution λ_x (as a vector labeled by x) by applying the matrix D (where $D_{xy} = |V_{xy}|^2$).

In general, we say that D is a stochastic map taking X to Y. if D has nonnegative entries with columns sum to 1.

Here, the rows of D also sum to 1, and we call it doubly stochastic. It is known to be entropy nondecreasing.

Therefore $S(\rho) = H(X) \leq H(Y)$ (meas outcomes are more random than the prep)

Properties of $S(A)$, $S(A|B)$, $S(A:B)$:

Like classical?

$$0. S(\rho) = S(U\rho U^\dagger), \quad S(A:B)_\rho = S(A:B)_{U\otimes V \rho U^\dagger \otimes V^\dagger}$$

$$1. 0 \leq S(A) \leq \log(\dim A) \quad Y$$

$$2. S(AB) \leq S(A) + S(B) \quad [\text{subadditivity}] \quad Y$$

$$\text{equiv to } S(A:B) \geq 0$$

$$\text{equiv to } S(A|B) \leq S(A)$$

$$"=" \text{ iff } \rho_{AB} = \rho_A \otimes \rho_B \text{ product state}$$

Proof: NC511. Def $S(\rho||\sigma)$
 $:= \text{tr}(\rho \log \rho) - \text{tr}(\rho \log \sigma)$
 Show $S(\rho||\sigma) \geq 0$ and
 $S(A:B)_\rho = S(\rho_{AB}||\rho_A \otimes \rho_B)$

3. Let τ_1, τ_2, \dots be states on the same system

and $\{p_k\}$ a distribution. Then,

$$S(\sum_k p_k \tau_k) \geq \sum_k p_k S(\tau_k) \quad Y$$

[entropy of the average \geq average entropy]

$$\text{Why: consider } \rho = \sum_k p_k \tau_k \otimes |k\rangle\langle k|.$$

$$S(AB) = H(p) + \sum_k p_k S(\tau_k)$$

$$S(A) = S(\sum_k p_k \tau_k), \quad S(B) = H(p). \text{ Follows from 2.}$$

Properties in the quantum setting:

Like classical
analogue?

$$4. S(A|B) \geq 0 \text{ or } S(A|B) \leq 0 \quad N$$

$$5. S(AB) = S(B) + S(A|B) \quad [\text{by def}] \quad Y$$

$$6. S(AB) \geq S(B) \text{ or } S(AB) \leq S(B)$$

7. Strong subadditivity (for any tripartite state on ABC)

$$S(C) + S(ABC) \leq S(AC) + S(BC) \quad Y$$

$$\text{equiv to } S(A|BC) \leq S(A|B) \quad \begin{array}{l} \text{conditioning reduces} \\ \text{conditional entropy} \end{array}$$

$$\text{equiv to } S(A:B|C) \geq 0 \quad \text{nonnegativity of conditional QMI}$$

$$\text{equiv to } S(A:B) \leq S(A:BC) \quad \text{local discarding cannot } \uparrow \text{ QMI}$$

$$\text{equiv to } S(A:B)_{(I \otimes E)(\rho)} \leq S(A:B)_\rho \quad \forall \text{ TCP } E \quad \begin{array}{l} \text{local processing} \\ \text{cannot } \uparrow \text{ QMI} \end{array}$$

Proof of SSA is very involved, see N&C 11.4.1.

7. Strong subadditivity

$$S(C) + S(ABC) \leq S(AC) + S(BC)$$

$$(i) \text{ equiv to } S(A|BC) \leq S(A|B)$$

$$(\text{so, } S(A:C|B) := S(A|B) - S(A|BC) \geq 0)$$

$$(ii) \text{ equiv to } S(A:B) \leq S(A:BC)$$

$$(iv) \text{ equiv to } S(A:B)_{(I \otimes E)(\rho)} \leq S(A:B)_\rho$$

Proof of equivalences:

$$(i) S(A|BC) = S(ABC) - S(BC), \quad S(A|B) = S(AB) - S(B)$$

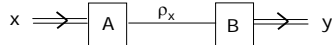
$$(ii) S(A:B) = S(A) + S(B) - S(AB), \quad S(A:BC) = S(A) + S(BC) - S(ABC)$$

(iii) recall any TCP map E can be realized by an isometry $B \rightarrow B'E$ where E is a suitable environment initially in a pure state, followed by discarding the environment.

The von Neumann entropy is invariant under a unitary change of basis. Thus $S(A:B) = S(A:B'E)$. By (ii), $S(A:B'E) \geq S(A:B)$. Conversely, discarding is a TCP map.

How much info can we learn about a quantum state by measuring it?

Given an ensemble $\mathcal{E} = \{p_x, \rho_x\}$, consider a game:



Alice draws x w.p. $p(x)$, prepares ρ_x and sends to Bob.

Bob performs meas \mathcal{M} on ρ_x with operators $\{M_y\}$ ($\sum_y M_y = I$).

Probability to obtain outcome y if state is ρ_x :

$$p(y|x) = \text{tr}(M_y \rho_x)$$

Joint distribution $p(xy) = p(y|x) p(x)$

Classical mutual info $I(X:Y)$ quantifies the information on which state X given by the measurement outcome Y

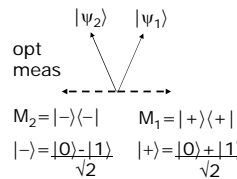
Def: $I_{\text{acc}}(\mathcal{E}) = \max_{\mathcal{M}} I(X:Y)$ [accessible info of \mathcal{E}]

I_{acc} is a natural quantity to define but difficult to compute.

Examples (proof of optimality of meas left as Ex/HW):

e.g.1 $\rho_1 = |\psi_1\rangle\langle\psi_1|$ for $|\psi_1\rangle = \cos \theta |0\rangle + \sin \theta |1\rangle$
 $\rho_2 = |\psi_2\rangle\langle\psi_2|$ for $|\psi_2\rangle = \cos \theta |0\rangle - \sin \theta |1\rangle$

drawn with $p_1 = p_2 = 1/2$



$$p(1|1) = 1/2 (\cos \theta + \sin \theta)^2 = \alpha$$

$$p(2|1) = 1/2 (\cos \theta - \sin \theta)^2 = 1 - \alpha$$

$$p(1|2) = 1/2 (\cos \theta - \sin \theta)^2 = 1 - \alpha$$

$$p(2|2) = 1/2 (\cos \theta + \sin \theta)^2 = \alpha$$

$$p(1) = p(2) = 1/2$$

$$H(Y|X) = H(\alpha), H(Y) = 1$$

$$I_{\text{acc}} = I(X:Y)$$

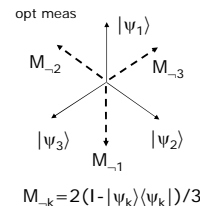
$$= H(Y) - H(Y|X) = 1 - H(\alpha)$$

$$\leq H(X) = 1.$$

Work for any ensemble with 2 pure states

e.g.2 $\rho_1 = |\psi_1\rangle\langle\psi_1|$ for $|\psi_1\rangle = |0\rangle$
 $\rho_2 = |\psi_2\rangle\langle\psi_2|$ for $|\psi_2\rangle = \cos \pi/3 |0\rangle + \sin \pi/3 |1\rangle$
 $\rho_3 = |\psi_3\rangle\langle\psi_3|$ for $|\psi_3\rangle = \cos \pi/3 |0\rangle - \sin \pi/3 |1\rangle$

drawn with $p_1 = p_2 = p_3 = 1/3$



$$H(X|Y) = 1$$

(each measurement outcomes rules out 1-out-of-3 states)

$$H(X) = \log 3$$

$$I_{\text{acc}} = I(X:Y)$$

$$= H(X) - H(X|Y) = (\log 3) - 1$$

do it however way is easier ≈ 0.5850

e.g.3 $\rho_x = |x\rangle\langle x|$ for $x = 0, 1, \dots, n-1$ and $\rho_{x+n} = U|x\rangle\langle x|U^\dagger$ When $n=2$, these are the 4 BB84 states
 $U = \text{fourier transform}$

Each state drawn with uniform probability $1/2n$.

(x is encoded in the computational or conjugate basis w.p. $1/2$ each)

Optimal measurement turns out to be $M_y = 1/2 \rho_y$ i.e. randomly measure in one of the two possible bases

Let T denote Bob's entire data set, where T is the coin toss specifying his measurement basis, and Y is the outcome of that measurement.

With prob $1/2$, Bob's random basis equals the actual one, giving $Y=X$, so, $I(X:Y|t \text{ correct}) = \log n$. With prob $1/2$, he measures in the "conjugate basis" so Y is random and independent of his quantum state (elaborate). So, $I(X:Y|t \text{ wrong}) = 0$. So, $I(X:Y) = 1/2 \log n$.

e.g.3 $\rho_x = |x\rangle\langle x|$ for $x = 0, 1, \dots, n-1$ and $\rho_{x+n} = U|x\rangle\langle x|U^\dagger$ When $n=2$, these are the 4 BB84 states
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Optimal measurement turns out to be $M_y = 1/2 \rho_y$ i.e. randomly measure in one of the two possible bases

We happen to find an upper bound of $1/2 \log n$ for I_{acc} .

Note: if 1 more bit (which basis) is given to Bob, he can always make the correct measurement, and $I_{\text{acc}} = 1 + \log n$.

So, I_{acc} can increase by $1 + 1/2 \log n$ bits when the system size increases by 1 bit. Since the increment $>>$ the extra bit, it does not "carry" the increment, but rather "unlocks" it from the other $\log n$ qubits. (More on locking later.)

A lower bound to accessible information

Jozsa, Robb, Wootters 94

For a density matrix ρ in d dimensions with eigenvalues $\{\lambda_k\}$, define the "subentropy":

$$Q(\rho) = -\sum_{k=1}^d [\Pi_{l \neq k} \lambda_l / (\lambda_k - \lambda_l)] \lambda_k \log \lambda_k$$

For the ensemble $\mathcal{E} = \{p_x, \rho_x\}$, $I_{\text{acc}}(\mathcal{E}) \geq Q(\sum_x p_x \rho_x) - \sum_x p_x Q(\rho_x)$ achieved by meas in random basis

If ρ_x are pure and $I/d = \sum_x p_x \rho_x$ (an ensemble of pure state that averages to the maximally mixed state),

$$I_{\text{acc}} \geq \log(d) - (\log e)(1/2 + 1/3 + \dots + 1/d) \text{ (in bits)}$$

$$\text{For } d = 2, I_{\text{acc}} \geq 0.2787, \text{ for } d \rightarrow \infty, I_{\text{acc}} \geq \approx 0.60995.$$

An upper bound to accessible information

For the ensemble $\mathcal{E} = \{p_x, \rho_x\}$, define

$$\text{Holevo information } \chi(\mathcal{E}) = S(\sum_x p_x \rho_x) - \sum_x p_x S(\rho_x)$$

Theorem: $I_{\text{acc}}(\mathcal{E}) \leq \chi(\mathcal{E})$

Proof:

The ensemble can be represented by the "CQ" state

$$\tau_{XQ} = \sum_x p_x |x\rangle\langle x| \otimes \rho_x$$

We interpret Alice as using the info x in system X to prepare the state ρ_x in system Q which is then transmitted to Bob.

Bob makes a measurement with POVM $\{M_y\}$ and outcome y stores in Y , and discards the system Q . The joint system is

$$\tau'_{XY} = \sum_x p_x |x\rangle\langle x| \otimes \sum_y \text{tr}(M_y \rho_x) |y\rangle\langle y|$$

Proof (ctd): $\tau_{XQY} = \sum_x p_x |x\rangle\langle x| \otimes \rho_x \otimes |0\rangle\langle 0|$

$$\tau'_{XQY} = \sum_x p_x |x\rangle\langle x| \otimes \sum_y M_y^{1/2} \rho_x M_y^{1/2} \otimes |y\rangle\langle y|$$

For any measurement by Bob:

$$S(X:Y)_{\tau'} \leq S(X:Q)_{\tau}$$

by monotonicity of QMI (since the state change is a TCP map on Bob's side).

For the optimal measurement, $S(X:Y)_{\tau'} = I_{\text{acc}}(\mathcal{E})$,

$$\begin{aligned} \text{whereas } S(X:Q)_{\tau} &= S(X) + S(Q) - S(XQ) \\ &= H(p) + S(\sum_x p_x \rho_x) - [H(p) + \sum_x p_x S(\rho_x)] \\ &= \chi(\mathcal{E}) \end{aligned}$$

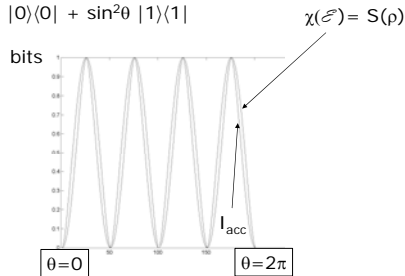
Therefore $I_{\text{acc}}(\mathcal{E}) \leq \chi(\mathcal{E})$.

Note a useful fact -- the Holevo information is the QMI of the "XQ" system defining the ensemble.

$$\begin{aligned} \text{e.g.1 } \rho_1 &= |\psi_1\rangle\langle\psi_1| \text{ for } |\psi_1\rangle = \cos\theta |0\rangle + \sin\theta |1\rangle \\ \rho_2 &= |\psi_2\rangle\langle\psi_2| \text{ for } |\psi_2\rangle = \cos\theta |0\rangle - \sin\theta |1\rangle \end{aligned}$$

drawn with $p_1 = p_2 = 1/2$

$$\rho = \cos^2\theta |0\rangle\langle 0| + \sin^2\theta |1\rangle\langle 1|$$



$$\begin{aligned} \text{e.g.2 } \rho_1 &= |\psi_1\rangle\langle\psi_1| \text{ for } |\psi_1\rangle = |0\rangle \\ \rho_2 &= |\psi_2\rangle\langle\psi_2| \text{ for } |\psi_2\rangle = \cos\pi/3 |0\rangle + \sin\pi/3 |1\rangle \\ \rho_3 &= |\psi_3\rangle\langle\psi_3| \text{ for } |\psi_3\rangle = \cos\pi/3 |0\rangle - \sin\pi/3 |1\rangle \end{aligned}$$

drawn with $p_1 = p_2 = p_3 = 1/3$

$$\rho = I/2, \chi = S(\rho) = 1, I_{\text{acc}} \approx 0.5850, Q(\rho) = 0.2787$$

A beautiful result is that, given 2 iid draws of this ensemble, the best joint measurement on the 4-dim system gives more than 2×0.5850 bits of information, so I_{acc} is not additive! Will learn later that I_{acc} on many copies is nearly $n\chi$.

$$\begin{aligned} \text{e.g.3 } \rho_x &= |x\rangle\langle x| \text{ for } x = 0, 1, \dots, n-1 \text{ and} \\ \rho_{x+n} &= U|x\rangle\langle x|U^\dagger \quad U = \text{fourier transform} \end{aligned}$$

Each state drawn with uniform probability $1/2n$.

$$\rho = I/n, \chi = S(\rho) = \log n, I_{\text{acc}} = 1/2 \log n, Q(\rho) \approx 0.6 \text{ for large } n$$

Note that in general, there are many many 1-qubit states, and to specify one such state takes many bits.

Preparing the quantum state (and not knowing the classical label) less than $S(I/2) = 1$ bit of info can be extracted.

It is highly irreversible.

Holevo's bound also says that we cannot use 1 qbit cannot transmit more one 1 bit of data.