Assignment 4 MATH 667 Quantum Information Theory

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1 Problem 1

Problem 1. Let $\mathcal{E}: \mathcal{H}_n \to \mathcal{H}_m$ be a linear map from the space of $n \times n$ Hermitian matrices to the space of $m \times m$ Hermitian matrices. The dual map $\mathcal{E}^*: \mathcal{H}_m \to \mathcal{H}_n$ is the linear map defined by the relation

$$\operatorname{Tr}[\rho \mathcal{E}^*(\sigma)] = \operatorname{Tr}[\mathcal{E}(\rho)\sigma] \quad \forall \rho \in \mathcal{H}_n \text{ and } \sigma \in \mathcal{H}_m.$$

- (a) Show that if \mathcal{E} is a CP map then \mathcal{E}^* is also a CP map.
- (b) Show that if \mathcal{E} is TP then \mathcal{E}^* is unital (i.e. show that $\mathcal{E}^*(I_m) = I_n$).
- (c) Let $\rho, \sigma \in \mathcal{H}_{n,+,1}$ be two density matrices. Show that $\sigma \prec \rho$ (i.e. the vector of eigenvalues of σ is majorized by that of ρ) if and only if there exists a unital CPTP map $\mathcal{E}: \mathcal{H}_n \to \mathcal{H}_n$ such that $\sigma = \mathcal{E}(\rho)$.

Solution. In the following, we use $\mathrm{id}_{\mathrm{H}_{\ell}}$ to denote the identity mapping on H_{ℓ} for some integer ℓ . Recall that a linear map $\Lambda:\mathrm{H}_n\to\mathrm{H}_m$ is completely positive if and only if the map $\Lambda\otimes\mathrm{id}_{\mathrm{H}_{\ell}}$ is positivity preserving for some integer $\ell\geq n$.

(a) Proof. Suppose that \mathcal{E} is a completely positive map, let $\ell = \max\{m, n\}$ and let $P \in H_m \otimes H_\ell$ be a positive operator. We need to show that $(\mathcal{E}^* \otimes \mathrm{id}_{H_\ell})(P) \in H_n \otimes H_\ell$ is a positive operator. Recall that an operator $A \in H_n \otimes H_\ell$ is positive if and only if it holds that $\mathrm{Tr}[AQ] \geq 0$ for all positive operators Q. Let $Q \in H_n \otimes H_\ell$ be a positive, then

$$\operatorname{Tr}[Q(\mathcal{E}^* \otimes \operatorname{id}_{H_{\ell}})(P)] = \operatorname{Tr}[(\mathcal{E} \otimes \operatorname{id}_{H_{\ell}})(Q)P] > 0$$

where $(\mathcal{E} \otimes \mathrm{id}_{H_{\ell}})(Q)$ is positive by the complete positivity of \mathcal{E} . Hence \mathcal{E}^* is completely positive. \square

(b) *Proof.* Suppose that \mathcal{E} is trace preserving. Note that the identity matrix I_n is the only operator $A \in \mathcal{H}_n$ with the property that $\langle \psi | A | \psi \rangle = 1$ holds for all unit vectors $|\psi\rangle \in \mathbb{C}^n$. Let $|\psi\rangle \in \mathbb{C}^n$ be an arbitrary unit vector. Then

$$\langle \psi | \mathcal{E}^*(I_m) | \psi \rangle = \text{Tr}[|\psi\rangle\langle\psi|\mathcal{E}^*(I_m)] = \text{Tr}[\mathcal{E}(|\psi\rangle\langle\psi|)I_m] = \text{Tr}[\mathcal{E}(|\psi\rangle\langle\psi|)] = \text{Tr}[|\psi\rangle\langle\psi|] = 1$$

since \mathcal{E} is trace-preserving. It follows that $\mathcal{E}^*(I_m) = I_n$, as desired.

(c) Proof. Let $\rho, \sigma \in \mathcal{H}_{n,+,1}$ be two density matrices and let $\vec{\lambda}$ and $\vec{\mu}$ be the vectors of the eigenvalues of ρ and σ respectively. There exist orthonormal bases $\{|u_i\rangle\}$ and $\{|v_i\rangle\}$ of \mathbb{C}^n such that

$$\rho = \sum_{i} \lambda_{i} |u_{i}\rangle\langle u_{i}| \quad \text{and} \quad \sigma = \sum_{i} \mu_{i} |v_{i}\rangle\langle v_{i}|.$$

Furthermore, define unitary operators $U = \sum_i |u_i\rangle\langle i|$ and $V = \sum_i |v_i\rangle\langle i|$ such that $\rho = U \operatorname{diag}(\vec{\lambda})U^*$ and $\sigma = V \operatorname{diag}(\vec{\mu})V^*$.

We first suppose that $\sigma \prec \rho$, that is $\vec{\mu} \prec \vec{\lambda}$. By assumption, there exists a doubly stochastic matrix D such that $\vec{\mu} = D\vec{\lambda}$. Since D is doubly stochastic, it can be written as a convex combination of permutation matrices

$$D = \sum_{\pi} p_{\pi} P_{\pi},$$

where the sum is taken over all permutations π of n elements, P_{π} is the permutation operator corresponding to π defined by

$$P_{\pi} = \sum_{i=1}^{n} |\pi(i)\rangle\langle i|,$$

and each $p_{\pi} \geq 0$ is a probability such that $\sum_{\pi} p_{\pi} = 1$. Note that $P_{\pi}^* = P_{\pi}^{-1}$ and thus $P_{\pi}^* P_{\pi} = P_{\pi} P_{\pi}^* = I_n$ holds for every permutation π . Therefore

$$\sum_{\pi} p_{\pi} P_{\pi}^* P_{\pi} = \sum_{\pi} p_{\pi} P_{\pi} P_{\pi}^* = \sum_{\pi} p_{\pi} I_n = I_n.$$

Consider the collection of operators $\{K_{\pi}\}$ where $K_{\pi} = \sqrt{p_{\pi}}VP_{\pi}^{*}U^{*}$ for each π . We see that

$$\sum_{\pi} K_{\pi}^* K_{\pi} = \sum_{\pi} p_{\pi} U P_{\pi} V^* V P_{\pi}^* U^* = U \left(\sum_{\pi} p_{\pi} P_{\pi} P_{\pi}^* \right) U^* = U I_n U^* = I_n$$

and thus the collection $\{K_{\pi}\}$ is a Kraus representation for the channel $\mathcal{E}: \mathcal{H}_n \to \mathcal{H}_n$ defined by

$$\mathcal{E}(A) = \sum_{\pi} K_{\pi} A K_{\pi}^* = \sum_{\pi} p_{\pi} V P_{\pi}^* U^* A U P_{\pi} V$$

for all $A \in \mathcal{H}_n$. Note that \mathcal{E} is unital. Indeed, the dual map \mathcal{E}^* can be given by $\mathcal{E}^*(A) = \sum_{\pi} K_{\pi}^* A K_{\pi}$ for all operators A and

$$\sum_{\pi} K_{\pi} K_{\pi}^{*} = \sum_{\pi} p_{\pi} V P_{\pi}^{*} U^{*} U P_{\pi} V^{*} = V \left(\sum_{\pi} p_{\pi} P_{\pi}^{*} P_{\pi} \right) V^{*} = V I_{n} V^{*} = I_{n}.$$

Hence \mathcal{E}^* is also a CPTP map with Kraus operators $\{K_{\pi}^*\}$. Since \mathcal{E}^* is trace preserving and $(\mathcal{E}^*)^* = \mathcal{E}$, it follows from part (b) that \mathcal{E} is unital.

Finally, note that $P_{\pi}^* \operatorname{diag}(\vec{\lambda}) P_{\pi} = \operatorname{diag}(P_{\pi}\vec{\lambda})$ holds for every π , since

$$P_{\pi}^* \operatorname{diag}(\vec{\lambda}) P_{\pi} = \sum_{i,j,k} |i\rangle \langle \pi(i)|(\lambda_j|j\rangle \langle j|) |\pi(k)\rangle \langle k| = \sum_i \lambda_i |\pi(i)\rangle \langle \pi(i)| = \sum_i \lambda_{\pi^{-1}(i)} |i\rangle \langle i| = \operatorname{diag}(P_{\pi}\vec{\lambda})$$

and $P_{\pi}\vec{\lambda} = \sum_{i} \lambda_{i} |\pi(i)\rangle = \sum_{i} \lambda_{\pi^{-1}(i)} |i\rangle$. Now

$$\begin{split} \mathcal{E}(\rho) &= \sum_{\pi} p_{\pi} V P_{\pi}^* U^* \rho U P_{\pi} V^* = \sum_{\pi} p_{\pi} V P_{\pi}^* \operatorname{diag}(\vec{\lambda}) P_{\pi} V^* \\ &= \sum_{\pi} p_{\pi} V \operatorname{diag}(P_{\pi} \vec{\lambda}) V^* \\ &= V \operatorname{diag}\left(\sum_{\pi} p_{\pi} P_{\pi} \vec{\lambda}\right) V^* = V \operatorname{diag}(\vec{\mu}) V^* = \sigma, \end{split}$$

and thus $\mathcal{E}(\rho) = \sigma$ for the unital CPTP map \mathcal{E} .

Now suppose that there exists a unital CPTP map \mathcal{E} such that $\sigma = \mathcal{E}(\rho)$. Define an $n \times n$ matrix D whose elements are given by

$$D_{ij} = \text{Tr}[|v_i\rangle\langle v_i|\mathcal{E}(|u_j\rangle\langle u_j|)].$$

Note that $\sum_i |u_i\rangle\langle u_i| = \sum_i |v_i\rangle\langle v_i| = I_n$, since both $\{|u_j\rangle\}$ and $\{|v_i\rangle\}$ are orthonormal bases. All of the columns and rows of D each sum to 1, since

$$\sum_{i} D_{ij} = \text{Tr}[I_n \mathcal{E}(|u_j\rangle\langle u_j|)] = \text{Tr}[\mathcal{E}(|u_j\rangle\langle u_j|)]] = 1$$

holds for all j by the fact that $\mathcal E$ is trace-preserving and

$$\sum_{i} D_{ij} = \text{Tr}[|v_i\rangle\langle v_i|\mathcal{E}(I_n)] = \text{Tr}[|v_i\rangle\langle v_i|] = 1$$

holds for all i by the fact that \mathcal{E} is unital. Furthermore, each D_{ij} is nonnegative, since $\mathcal{E}(|u_j\rangle\langle u_j|)$ is a positive operator by positivity of \mathcal{E} , and thus

$$D_{ij} = \langle v_i | \mathcal{E}(|u_i\rangle\langle u_j|) | v_i \rangle \ge 0$$

for all i and j. It follows that D is a doubly stochastic matrix.

We now show that $\vec{\mu} = D\vec{\lambda}$. The i^{th} entry of $D\vec{\lambda}$ is

$$(D\vec{\lambda})_i = \sum_j \text{Tr}[|v_i\rangle\langle v_i|\mathcal{E}(|u_j\rangle\langle u_j|)]\lambda_j$$

$$= \text{Tr}\Big[|v_i\rangle\langle v_i|\mathcal{E}\Big(\underbrace{\sum_j \lambda_j |u_j\rangle\langle u_j|}_{\rho}\Big)\Big] = \text{Tr}[|v_i\rangle\langle v_i|\sigma] = \langle v_i|\sigma|v_i\rangle = \mu_i,$$

and thus $\vec{\mu} = D\vec{\lambda}$ for a doubly stochastic matrix D. It follows that $\vec{\mu} \prec \vec{\lambda}$ and thus $\sigma \prec \rho$, as desired. \Box

2 Problem 2

Problem 2. Find necessary and sufficient conditions for which the following equality holds:

$$S(\rho^{\mathsf{AB}}) = |S(\rho^{\mathsf{A}}) - S(\rho^{\mathsf{B}})|. \tag{2.1}$$

Give an example.

Solution. The necessary and sufficient conditions for equality come from the following observation together with Proposition 1. The problem can be broken down into two sub-statements:

(i)
$$S(\rho^{AB}) = S(\rho^{B}) - S(\rho^{A})$$
 if and only if $\rho^{AR} = \rho^{A} \otimes \rho^{R}$ holds for all purifications ρ^{ABR} of ρ^{AB} , and

(ii)
$$S(\rho^{AB}) = S(\rho^{A}) - S(\rho^{B})$$
 if and only if $\rho^{BR} = \rho^{B} \otimes \rho^{R}$ holds for all purifications ρ^{ABR} of ρ^{AB} .

The following proposition proves only part (i), but flipping the A and B yields part (ii). Note that $S(\rho^{AB})$ must be nonnegative. Hence equality in (2.1) holds if and only if the condition in either (i) or (ii) holds.

Proposition 1. Let ρ^{AB} be a bipartite state. Then the equality $S(\rho^{AB}) = S(\rho^{B}) - S(\rho^{A})$ holds if and only if it holds that $\rho^{AR} = \rho^{A} \otimes \rho^{R}$ for all possible purifications ρ^{ABR} of ρ^{AB} .

Proof. Let ρ^{ABR} be a purification of ρ^{AB} such that $\rho^{ABR} = |\psi\rangle\langle\psi|^{ABR}$ for a pure state vector $|\psi\rangle^{ABR}$ and

$$\rho^{\mathsf{AB}} = \mathrm{Tr}_{\mathsf{R}} \, \rho^{\mathsf{ABR}}.$$

Suppose that $\rho^{AR} = \rho^A \otimes \rho^R$. Recall that the von Neumann entropy is sub-additive, $S(\rho^{AR}) \leq S(\rho^A) + S(\rho^R)$ with equality if and only if $\rho^{AR} = \rho^A \otimes \rho^R$ (i.e., systems A and R are uncorrelated). It follows that

$$S(\rho^{\mathsf{AR}}) = S(\rho^{\mathsf{A}}) + S(\rho^{\mathsf{R}}) \tag{2.2}$$

from the assumption. Since the state $\rho^{ABR} = |\psi\rangle\langle\psi|^{ABR}$ is pure, it holds that

$$S(\rho^{\mathsf{R}}) = S(\rho^{\mathsf{AB}})$$
 and $S(\rho^{\mathsf{B}}) = S(\rho^{\mathsf{AR}}).$ (2.3)

Putting together equations (2.2) and (2.3) yields the equality $S(\rho^{AB}) = S(\rho^{B}) - S(\rho^{A})$.

For the converse, the exact same argument works in reverse. That is, if we suppose that $S(\rho^{AB}) = S(\rho^B) - S(\rho^A)$, we can make the same replacements as above to yield the equality $S(\rho^{AR}) = S(\rho^A) + S(\rho^R)$ for any possible purification ρ^{ABR} . This equality occurs if and only if $\rho^{AR} = \rho^A \otimes \rho^R$, as desired.

Example. The "trivial" examples of states ρ^{AB} that satisfy this equality in (2.1) are those for which either

- ρ^{AB} is pure, in which case $S(\rho^{\mathsf{AB}}) = 0$ and $S(\rho^{\mathsf{A}}) = S(\rho^{\mathsf{B}})$, or
- $\rho^{AB} = \rho^A \otimes \rho^B$ and at least one of ρ^A or ρ^B is pure, in which case $S(\rho^{AB}) = S(\rho^A)$ if ρ^B is pure or $S(\rho^{AB}) = S(\rho^B)$ if ρ^A is pure.

We can construct a non-trivial example of a state ρ^{AB} that satisfies the equality in (2.1) in the following. Consider the pure state vectors $|u\rangle$ and $|v\rangle$ in $\mathbb{C}^2\otimes\mathbb{C}^4$ defined by

$$|u\rangle^{\mathsf{AB}} = \frac{1}{\sqrt{2}}(|00\rangle^{\mathsf{AB}} + |11\rangle^{\mathsf{AB}})$$
 and $|v\rangle^{\mathsf{AB}} = \frac{1}{\sqrt{2}}(|02\rangle^{\mathsf{AB}} + |13\rangle^{\mathsf{AB}}).$

Note that $\langle u|v\rangle=0$. Define the following mixed state (which is certainly not pure)

$$\rho^{\mathsf{AB}} = \frac{1}{2} |u\rangle\langle u| + \frac{1}{2} |v\rangle\langle v|$$

and note that $S(\rho^{\mathsf{AB}}) = -\frac{1}{2}\log\frac{1}{2} - \frac{1}{2}\log\frac{1}{2} = \log 2 = 1$. The reduced density operators on systems A and B are

$$\rho^{\mathsf{A}} = \frac{1}{2} \big(|0\rangle\langle 0| + |1\rangle\langle 1| \big) \qquad \text{and} \qquad \rho^{\mathsf{B}} = \frac{1}{4} \big(|0\rangle\langle 0| + |1\rangle\langle 1| + |2\rangle\langle 2| + |3\rangle\langle 3| \big),$$

both of which are clearly not pure. The entropies of these reduced density operators are $S(\rho^{\mathsf{A}}) = \log 2 = 1$ and $S(\rho^{\mathsf{B}}) = \log 4 = 2$. We see that

$$|S(\rho^{\mathsf{A}}) - S(\rho^{\mathsf{B}})| = |1 - 2| = 1 = S(\rho^{\mathsf{AB}}),$$

as desired.

3 Problem 3

Problem 3. Prove the strong sub-additivity of the Shannon entropy: For three random variables X, Y, Z,

$$H(X,Y,Z) + H(Y) \le H(X,Y) + H(Y,Z)$$

where H is the Shannon entropy.

Solution. Proof. Define the mutual information of two variables X and Y as

$$I(X:Y) = H(X) + H(Y) - H(X,Y).$$

From the sub-additivity of the Shannon entropy, $H(X,Y) \leq H(X) + H(Y)$, it is clear that $I(X:Y) \geq 0$. Define the conditional mutual information as

$$I(X \colon Z \mid Y) = \sum_{y} p(y)I(X|_{Y=y} \colon Z|_{Y=y}).$$

Since both p(y) and $I(X|_{Y=y}: Z|_{Y=y})$ are nonnegative for every y, it follows that $I(X: Z|Y) \ge 0$. Recall that $p(x|Y=y) = \frac{p(x,y)}{p(y)}$. By definition of the mutual information, we have

$$\begin{split} p(y)I(X|_{Y=y}\colon Z|_{Y=y}) &= p(y)H(X|_{Y=y}) + p(y)H(Z|_{Y=y}) - p(y)H(X,Z|_{Y=y}) \\ &= -\sum_x p(x,y)\log\frac{p(x,y)}{p(y)} - \sum_z p(z,y)\log\frac{p(z,y)}{p(y)} + \sum_{x,z} p(x,y,z)\log\frac{p(x,y,z)}{p(y)} \\ &= -\sum_x p(x,y)\log p(x,y) - H(Y) - \sum_z p(z,y)\log p(z,y) - H(Y) \\ &\qquad + \sum_{x,z} p(x,y,z)\log p(x,y,z) + H(Y) \\ &= -\sum_x p(x,y)\log p(x,y) - \sum_z p(z,y)\log p(z,y) + \sum_{x,z} p(x,y,z)\log p(x,y,z) - H(Y). \end{split}$$

By taking the sum over all y's in the definition of I(X:Z|Y), we have

$$I(X: Z | Y) = \sum_{y} p(y)I(X|_{Y=y}: Z|_{Y=y})$$

= $H(X,Y) + H(Y,Z) - H(X,Y,Z) - H(Y).$

Since $I(X: Z | Y) \ge 0$, from this it follows that

$$H(X,Y) + H(Y,Z) - H(X,Y,Z) - H(Y) \ge 0$$

and thus $H(X,Y,Z) + H(Y) \le H(X,Y) + H(Y,Z)$ as desired.

4 Problem 4

Problem 4. Consider an i.i.d. source characterized by a random variable X with alphabet $x \in \{1, \ldots, d\} = \mathcal{X}$ corresponding to probability p(x) > 0. Consider a sequence of size n, denoted as $x^n = (x_1, \ldots, x_n) \in \mathcal{X}^n$. The empirical distribution of the sequence x^n is defined by

$$q_{x^n}(x) = \frac{1}{n} N(x|x^n)$$

where $N(x|x^n)$ is the number of times the symbol $x \in \{1, \ldots, d\}$ appears in x^n . For $\delta > 0$ denote

$$T(n,\delta) = \{x^n \in \mathcal{X}^n \mid |q_{x^n}(x) - p(x)| < \delta \ \forall x \in \mathcal{X}\}.$$

(a) Show that for any $\epsilon, \delta > 0$ and sufficiently large n

$$\Pr(T(n,\delta)) \ge 1 - \epsilon.$$

(b) Show that for any $\epsilon, \delta > 0$ and sufficiently large n

$$(1 - \epsilon)2^{n(H(X) - c\delta)} \le |T(n, \delta)| \le 2^{n(H(X) + c\delta)}$$

for some positive constant c.

(c) Show that if $x^n \in T(n, \delta)$ then

$$2^{-n(H(X)+c\delta)} \le p(x^n) \le 2^{-n(H(X)-c\delta)} \tag{4.1}$$

for some positive constant c, and $p(x^n) = p(x_1)p(x_2)\cdots p(x_n)$.

Solution. The idea for this solution is due to Wilde¹ and Yeung².

(a) Let $\epsilon, \delta > 0$. For each $x \in \mathcal{X}$, consider the i.i.d. indicator random variables $I_1(x), \ldots, I_k(x)$ obtained by sampling x_i and then setting

$$I_i(x) = \begin{cases} 1 \text{ if } x_i = x \\ 0 \text{ if } x_i \neq x \end{cases}$$

We can write $N(x|x^n)$ as

$$N(x|x^n) = \sum_{i=1}^n I_i(x).$$

Since $\Pr(I_i(x) = 1) = p(x)$, we have the expected values $E[I_i(x)] = p(x)$ for all $x \in \mathcal{X}$. By the weak law of large numbers, for every $a \in \mathcal{X}$, there is a sufficiently large n_a such that

$$\Pr\left(\left\{\left|q_{x^n}(a) - p(a)\right| \ge \delta\right\}\right) = \Pr\left(\left\{\left|\frac{1}{n}\sum_{i=1}^n I_i(a) - p(a)\right| \ge \delta\right\}\right) < \frac{\epsilon}{d}$$

$$(4.2)$$

holds for all $n > n_a$, where $d = |\mathcal{X}|$. Let $n_0 = \max\{n_a \mid a \in \mathcal{X}\}$. Then for all $n > n_0$, we have

$$\Pr\left(\left\{\left|q_{x^n}(a) - p(a)\right| \ge \delta \text{ for some } a \in \mathcal{X}\right\}\right) = \Pr\left(\left\{\left|\frac{1}{n}\sum_{i=1}^n I_i(a) - p(a)\right| \ge \delta \text{ for some } a \in \mathcal{X}\right\}\right)$$

$$= \Pr\left(\bigcup_{a \in \mathcal{X}} \left\{\left|\frac{1}{n}\sum_{i=1}^n I_i(a) - p(a)\right| \ge \delta\right\}\right)$$

¹See Section 14.7.1 - 14.7.3 in Classical to Quantum Shannon Theory by Mark Wilde.

²See Chapter 6 in *Information Theory and Network Coding* by Raymond Yeung.

$$\leq \sum_{a \in \mathcal{X}} \Pr\left(\left\{ \left| \frac{1}{n} \sum_{i=1}^{n} I_i(a) - p(a) \right| \geq \delta \right\} \right)$$

$$< \sum_{a \in \mathcal{X}} \frac{\epsilon}{d} = \epsilon,$$

where the first inequality follows from the union bound³ and the last inequality follows from (4.2). The probability of the complement of the above event is thus

$$\Pr\left(\left\{\left|q_{x^n}(a) - p(a)\right| < \delta \text{ for all } a \in \mathcal{X}\right\}\right) = 1 - \epsilon$$

for all $n > n_0$, as desired.

(b) We will prove part (b) from part (c). Hence, assuming part (c) is true, we can use the upper bound in (4.1) (i.e. $p(x^n) \le 2^{-n(H(X)-c\delta)}$) to find that

$$\Pr(T(n,\delta)) = \sum_{x^n \in T(n,\delta)} p(x^n) \le |T(n,\delta)| 2^{-n(H(X) - c\delta)}$$

for some positive constant c. For n sufficiently large, we have $1 - \epsilon \leq \Pr(T(n, \delta))$, and it follows from part (a) that

$$1 - \epsilon \le \Pr(T(n, \delta)) \le |T(n, \delta)| 2^{-n(H(X) - c\delta)}$$

Thus $(1-\epsilon)2^{n(H(X)-c\delta)} \leq |T(n,\delta)|$, which yields the desired lower bound. For the upper bound, we use the lower bound in (4.1) (i.e. $\leq 2^{-n(H(X)+c\delta)} \leq p(x^n)$). Now

$$|T(n,\delta)|2^{-n(H(X)+c\delta)} \le \sum_{x^n \in T(n,\delta)} p(x^n) = \Pr(T(n,\delta)) \le 1,$$

and rearranging yields the desired upper bound $|T(n,\delta)| \leq 2^{n(H(X)+c\delta)}$. (Note that this upper bound holds for all n.)

(c) If $p(a) \neq 0$ for every $a \in \mathcal{X}$, for any sequence x^n we can write

$$p(x^n) = \prod_{a \in \mathcal{X}} p(a)^{N(a|x^n)}.$$

If p(a) = 0 for any $a \in \mathcal{X}$, then we must consider the subset \mathcal{X}^+ of \mathcal{X} containing all a for which p(a) > 0. For n large enough, it must hold that

$$p(x^n) = \prod_{a \in \mathcal{X}^+} p(a)^{N(a|x^n)}.$$
(4.3)

for any $x^n \in T(n, \delta)$. That is, no symbols $a \in \mathcal{X}$ with probability p(a) = 0 can appear in typical sequences x^n if n is sufficiently large. Taking the logarithm of (4.3) yields

$$\log p(x^n) = \sum_{a \in \mathcal{X}^+} N(a|x^n) \log p(a)$$

$$= \sum_{a \in \mathcal{X}^+} \left(N(a|x^n) + np(a) - np(a) \right) \log p(a)$$

$$= n \sum_{a \in \mathcal{X}^+} p(a) \log p(a) + n \sum_{a \in \mathcal{X}^+} \left(\frac{1}{n} N(a|x^n) - p(a) \right) \log p(a)$$

³The union bound refers to the fact that $\Pr\left(\bigcup_i A_i\right) \leq \sum_i \Pr(A_i)$ for any set of events A_i .

$$= -n \left(H(X) + \sum_{a \in \mathcal{X}^+} \left(\frac{1}{n} N(a|x^n) - p(a) \right) \left(-\log p(a) \right) \right).$$

Since $x^n \in T(n, \delta)$, we know

$$\left| \frac{1}{n} N(a|x^n) - p(a) \right| \le \delta$$

which implies

$$\left| \sum_{a \in \mathcal{X}^+} \left(\frac{1}{n} N(a|x^n) - p(a) \right) \left(-\log p(a) \right) \right| \le \sum_{a \in \mathcal{X}^+} \left| \left(\frac{1}{n} N(a|x^n) - p(a) \right) \right| \left(-\log p(a) \right)$$

$$\le \delta \sum_{a \in \mathcal{X}^+} \left(-\log p(a) \right)$$

$$= \delta c$$

where we take c to be the positive constant

$$c = -\sum_{a \in \mathcal{X}^+} \log p(a).$$

From this we see that

$$-\delta c \le \sum_{a \in \mathcal{X}^+} \left(\frac{1}{n} N(a|x^n) - p(a) \right) \left(-\log p(a) \right) \le \delta c$$

and thus

$$-n(H(X) + \delta c) \le \underbrace{-n\left(H(X) + \sum_{a \in \mathcal{X}^+} \left(\frac{1}{n}N(a|x^n) - p(a)\right)\log p(a)\right)}_{\log p(x^n)} \le -n(H(X) - \delta c),$$

which yields

$$2^{-n(H(X)+\delta c)} \le p(x^n) \le 2^{-n(H(X)-\delta c)}$$

as desired.