

Short proofs of the Quantum Substate Theorem

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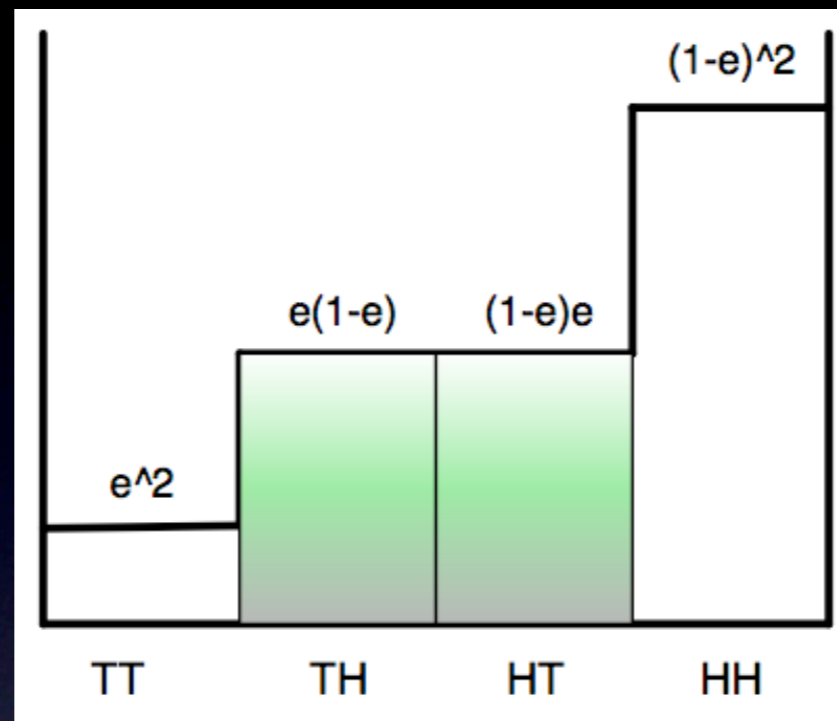
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Classic problem

- Given a biased coin c
 - $\Pr(c = H) = 1 - e$
 - $\Pr(c = T) = e$
- Can we generate a fair coin toss ?

Lemonade from lemons



Von Neumann: rejection sampling

- Toss c twice
- Repeat if HH or TT, else output result

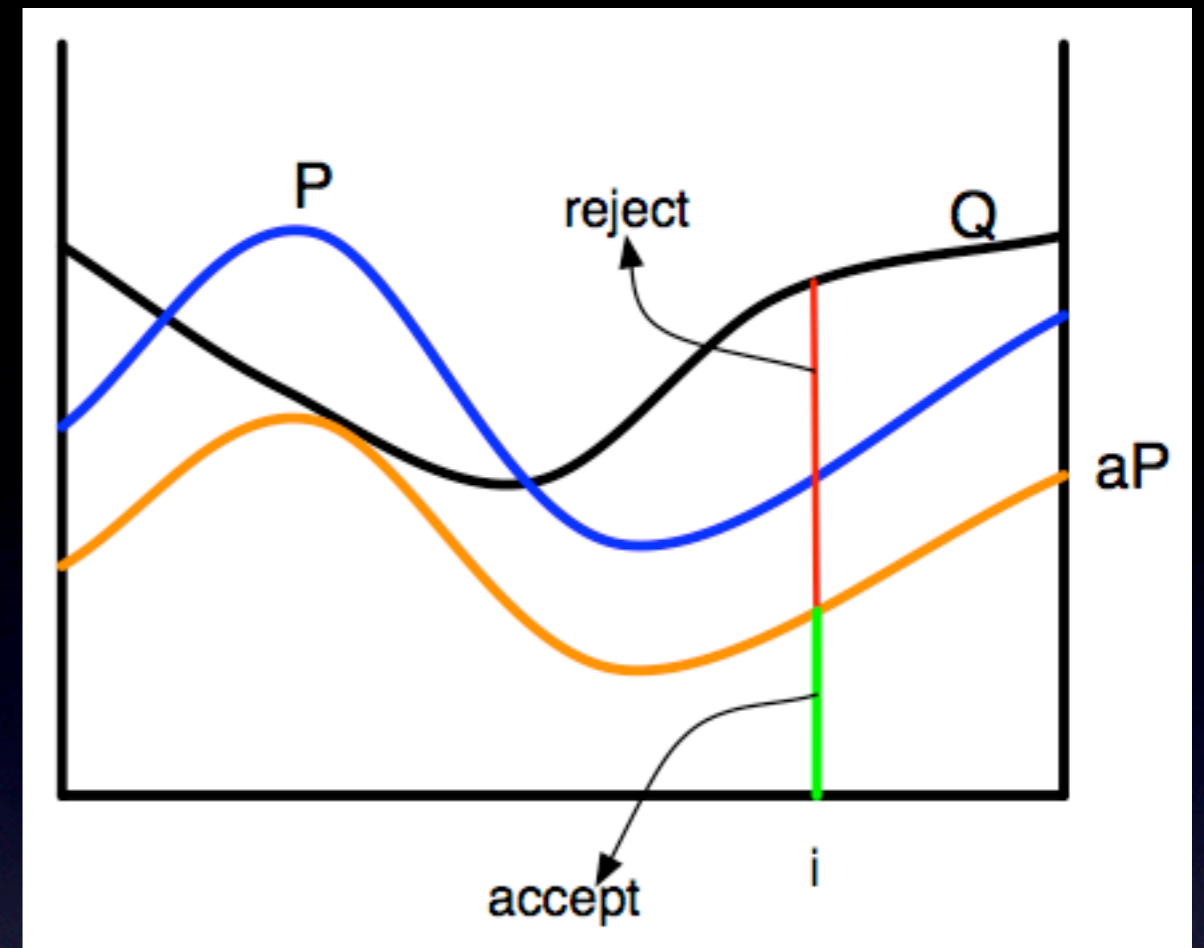
$$\Pr(\text{success in 1 trial}) = p = 2e(1-e)$$

$$E(\# \text{ trials for success}) = 1/p$$

Quantum substate theorem

- Say we are given a quantum state Q , but we wish to prepare state P
- The theorem gives a bound on the number of trials a quantum analogue of rejection sampling takes (formal statement later)
- Original proof by Jain, Radhakrishnan, and Sen (2002)
- Applications in cryptography, communication and information theory
- This talk: short, conceptually simple proof of the theorem
- Stronger statement, optimal up to a constant factor

Classical version



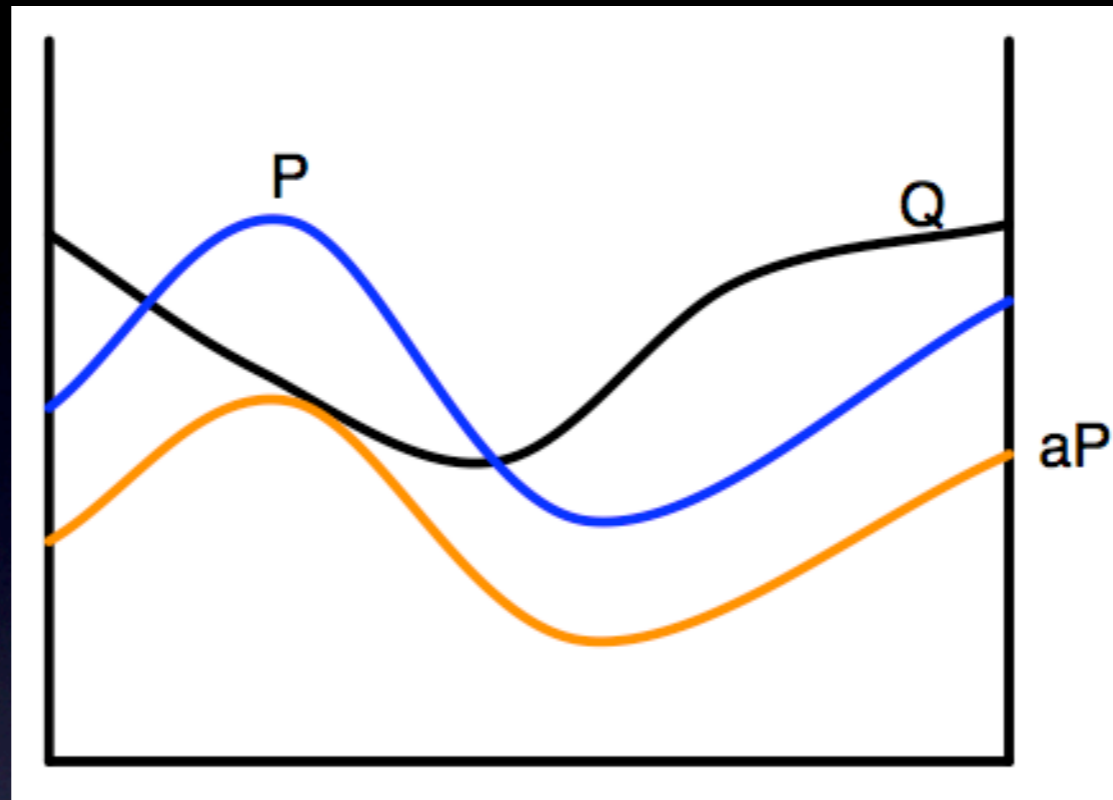
Rejection sampling

- Scale P so its graph is contained within Q : $aP \leq Q$
- We get a sample i from Q
- Throw a dart uniformly at random on the vertical line up to Q
- Repeat with new sample if dart above aP , else output i

$$\Pr(\text{success in 1 trial}) = a = \min_i q_i / p_i$$

$$E(\# \text{ trials for success}) = 1/a = \max_i p_i / q_i$$

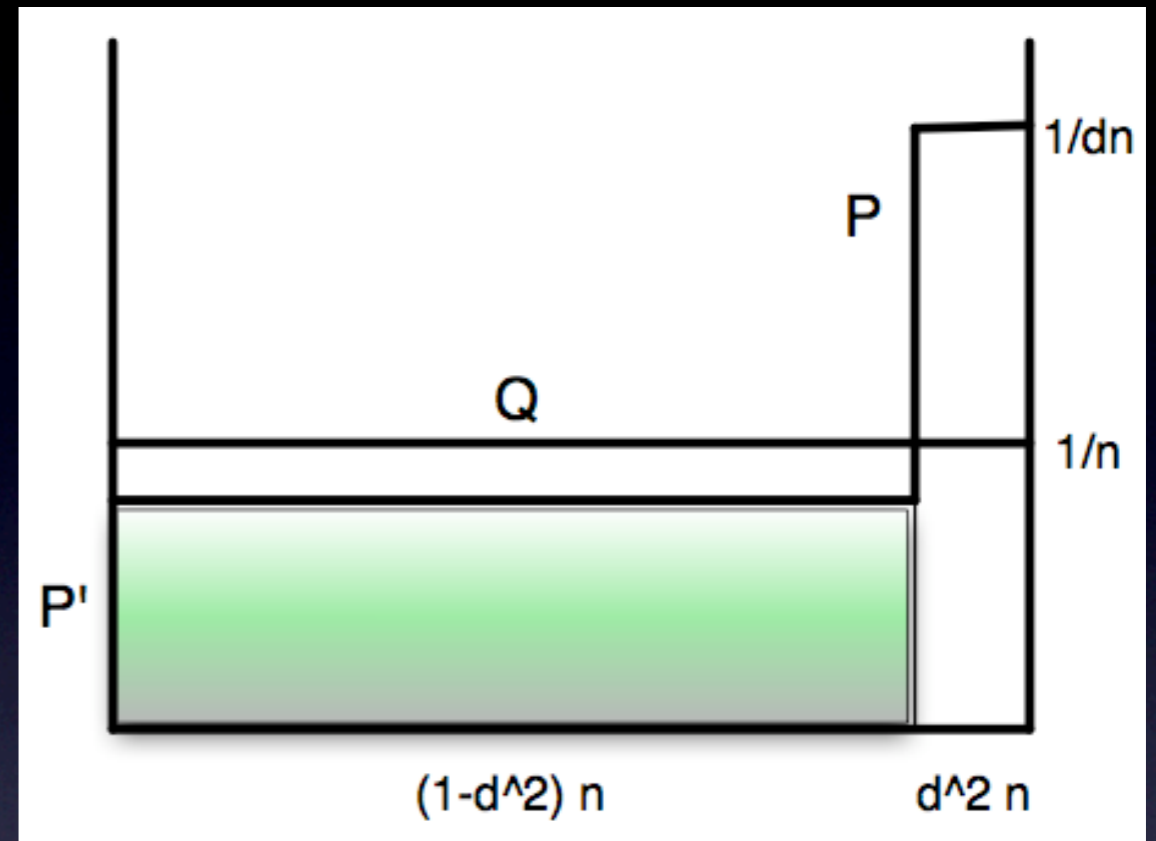
Relative min-entropy



- We say aP is a subdistribution of Q
- $E(\# \text{ trials for success}) = 1/a = \max_i p_i / q_i$
- Important measure of distance between distributions P, Q
- $S_\infty(P|Q) = \log_2(1/a) = \max_i \log_2(p_i / q_i)$
- Relative entropy: $S(P|Q) = \sum_i p_i \log_2(p_i / q_i) \leq S_\infty(P|Q)$

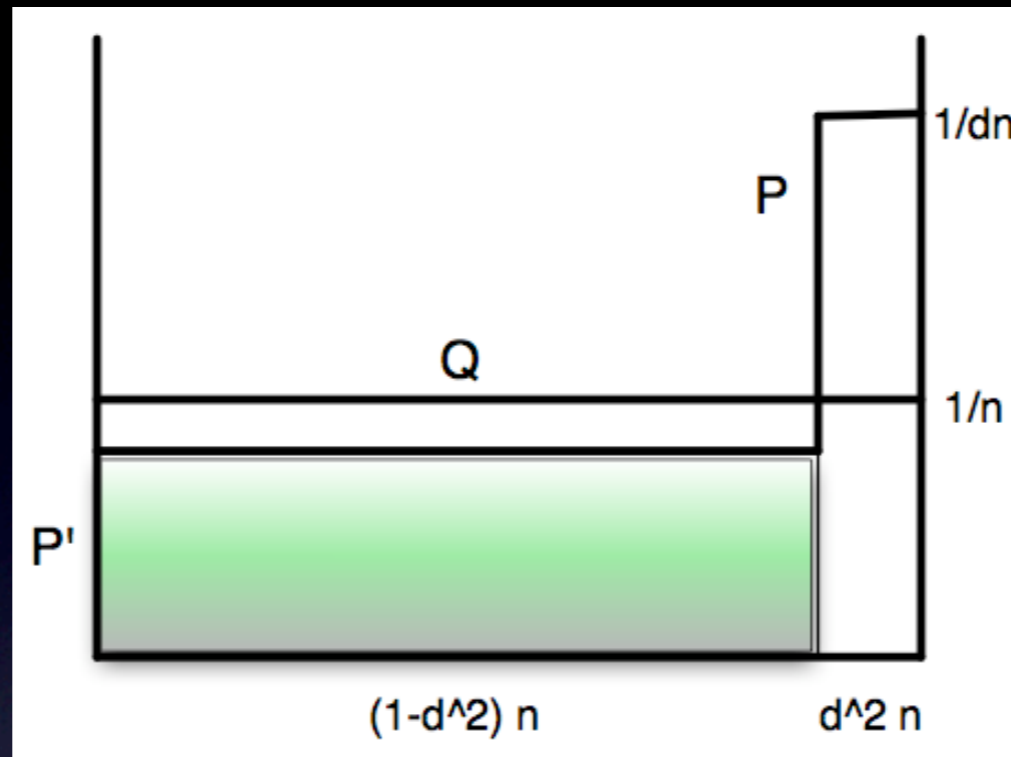
Approximate sampling

- Given Q often suffices to generate P' close to P
- Say, $|P' - P| \leq \epsilon$
(1/2 L_1 distance)



- $S_\infty(P|Q) = \log_2(1/d)$
- Let P' be uniform on the first $(1 - d^2) n$ points
- Then $|P' - P| \leq d$
- $S_\infty(P'|Q) = \log_2(1 / (1 - d^2)) \approx \text{const } d^2$

Smooth relative min-entropy



- Interested in P' ϵ -close to P , such that aP' is a subdistribution of Q and a is maximized
- $E(\# \text{ trials for success}) = 1/a$
- $S_{\infty}^{\epsilon}(P|Q) = \log_2 \min \{ 1/a : aP' \leq Q, |P' - P| \leq \epsilon \}$
 $= \log_2 \min \{ k : P' \leq kQ, |P' - P| \leq \epsilon \}$
- How do we estimate this quantity ?

Substate theorem [JRS'02]

Theorem

Suppose P, Q are probability distributions
with $\text{supp}(P) \subseteq \text{supp}(Q)$.

For every $\epsilon \in (0, 1)$ there is a distribution P' such that

$$|P' - P| \leq \epsilon, \text{ and}$$

$$P' \leq \left[2^{(s+1)/\epsilon} / (1-\epsilon) \right] Q,$$

where $s = S(P|Q)$.

$$\text{i.e., } S^\epsilon_\infty(P|Q) \leq (S(P|Q) + 1) / \epsilon + \log_2 1/(1 - \epsilon)$$

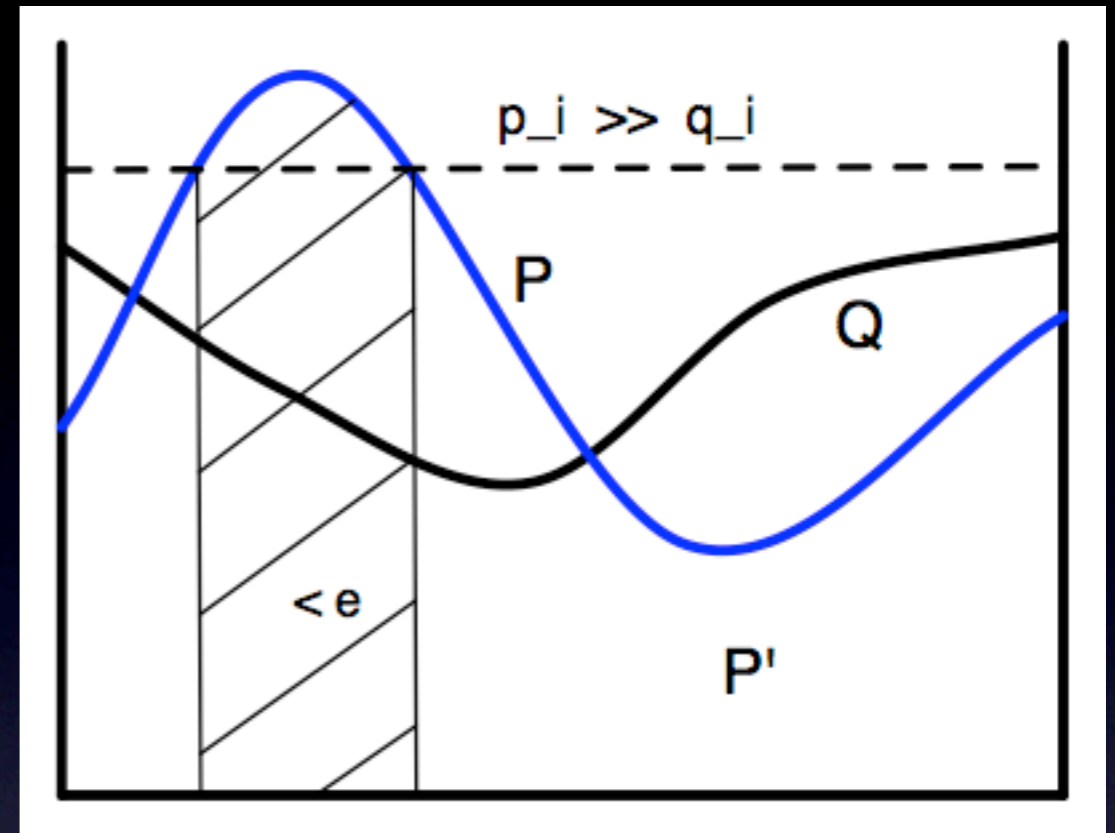
Recall: relative entropy $S(P|Q) = \sum_i p_i \log_2 (p_i / q_i)$.

Proof of Substate Theorem

Let $s = S(P|Q) = \sum_i p_i \log_2 (p_i / q_i)$

If p_i/q_i were at least 1, we could use the Markov inequality:

$$\Pr(\log_2 (p_i/q_i) \geq s/e) \leq e$$



The sum of negative terms $\sum_i p_i \log_2 (p_i / q_i)$ is at least -1 .

$$\Rightarrow \Pr(\log_2 (p_i/q_i) \geq (s+1)/e) \leq e$$

Let $P' = P$ conditioned on the complementary event.

We have $|P' - P| \leq e$, and $(1-e) P' \leq 2^{(s+1)/e} Q$ ■

Quantum substate theorem [JRS'02]

Theorem

quantum states

Suppose P, Q are probability distributions
with $\text{supp}(P) \subseteq \text{supp}(Q)$.

quantum state

For every $\epsilon \in (0, 1)$ there is a distribution P' such that

$$|P' - P| \leq \underline{\epsilon} \sqrt{\epsilon} \quad \text{and}$$

$$P' \leq [2^{(s+1)/\epsilon} / (1-\epsilon)] Q,$$

where $s = S(P|Q)$.

$$\text{i.e., } S^d_\infty(P|Q) \leq (S(P|Q) + 1) / \epsilon + \log_2 1/(1 - \epsilon)$$

where $d = \sqrt{\epsilon}$

Decoding this theorem

- What is a quantum state ?
- What is rejection sampling for quantum states ?
- What is the relative entropy of quantum states ?
- When are two quantum states close to each other ?

Quantum state

- A quantum state P is a positive semidefinite operator on \mathbb{C}^n with unit trace
- Simplest case: rank 1 operator
 - $P = v v^*$, v is called a *superposition*
- Rank > 1 : $P = \sum_i p_i v_i v_i^*$ (spectral decomposition)
 - may be viewed as a probability distribution (p_i) over the eigenvectors v_i
- Probability distribution over $\{1, \dots, n\}$ is a diagonal such operator: $v_i = e_i$, standard basis vectors

Rejection sampling



- Given quantum state Q , would instead like to prepare state P

- We say aP is a substate of Q if $aP \leq Q$, i.e.,

$$Q = aP + (1-a)Q', \quad \text{where } Q' \geq 0$$

- Several quantum variants of rejection sampling, all have success probability a
- Again, $E(\# \text{ trials for success}) = 1/a$

Relative min-entropy

- Important measure of distance between quantum states P, Q :
 - what is the maximum a such that $aP \leq Q$?
 - least E (# trials for success) = $1/a$
- $S_\infty(P|Q) = \log_2 \min \{ 1/a : aP \leq Q \}$
 $= \log_2 \min \{ k : P \leq kQ \}$

No simple expression for this in terms of P, Q

- Relative entropy: $S(P|Q) = \text{Tr } P (\log_2 P - \log_2 Q)$
- \log_2 is an operator monotone function
 - $\log_2 P \leq (\log_2 k) I + \log_2 Q$
- So $S(P|Q) \leq S_\infty(P|Q)$

Smooth relative min-entropy

- Suppose we can tolerate some error ϵ in generating the quantum state P from Q :
 - would like P' ϵ -close to P , such that aP' is a substate of Q and a is maximized
- Distance measure: induced by trace norm (Schatten 1-norm)
 - $|M| = (1/2) \text{Tr} (M^*M)^{1/2} = (1/2)$ sum of singular values
 - tells us how well two quantum states may be distinguished
- $S_{\infty}^{\epsilon}(P|Q) = \log_2 \min \{ 1/a : aP' \leq Q, |P' - P| \leq \epsilon \}$
 $= \log_2 \min \{ k : P' \leq kQ, |P' - P| \leq \epsilon \}$
- How do we estimate this quantity ?

Quantum substate theorem [JRS'02]

Theorem

Suppose P, Q are quantum states, $\text{supp}(P) \subseteq \text{supp}(Q)$.

For every $\epsilon \in (0,1)$ there is a quantum state P' such that

$$|P' - P| \leq \sqrt{\epsilon}, \quad \text{and}$$

$$P' \leq \left[2^{(s+1)/\epsilon} / (1-\epsilon) \right] Q,$$

where $s = S(P|Q)$.

$$\text{i.e., } S^d_\infty(P|Q) \leq (S(P|Q) + 1) / \epsilon + \log_2 1/(1 - \epsilon)$$

where $d = \sqrt{\epsilon}$

New proof [Jain, N'11]

- Key observation: smooth relative min-entropy is the logarithm of the following convex program (SDP) over variables P', k

$$\min k$$

such that

$$P' \leq kQ$$

$$|P' - P| \leq \epsilon$$

$$\text{Tr}(P') = 1$$

$$P' \geq 0$$

- Using strong duality, it suffices to bound the dual optimum
- Bound on dual is analogous to the substate theorem for distributions

First use of duality

$$A \leq B \quad \Leftrightarrow \quad v^*Av \leq v^*Bv \quad \text{for all } v \in \mathbb{C}^n$$

$$\Leftrightarrow \quad \text{Tr}(vv^*A) \leq \text{Tr}(vv^*B) \quad \text{for all } v \in \mathbb{C}^n$$

$$\Leftrightarrow \quad \text{Tr}(MA) \leq \text{Tr}(MB) \quad \text{for all } M \geq 0$$

First use of duality

Lemma:

Suppose P', Q are quantum states with $\text{supp}(P') \subseteq \text{supp}(Q)$.

Then

$$\begin{aligned} \min \{ k : P' \leq kQ \} \\ = \max \{ \text{Tr}(MP') : \text{Tr}(MQ) = 1, M \geq 0 \}. \end{aligned}$$

Proof:

$$P' \leq kQ \iff \text{Tr}(MP') \leq k \text{Tr}(MQ) \quad \text{for all } M \geq 0$$

$$\iff \text{Tr}(MP') \leq k \text{Tr}(MQ) \quad \text{for all } M \geq 0, \text{Tr}(MQ) \neq 0$$

$$\iff \text{Tr}(MP') \leq k \quad \text{for all } M \geq 0, \text{Tr}(MQ) = 1$$

(scale M by $\text{Tr}(MQ)$) ■

A min-max formulation

The convex program over variables P', k

$$\begin{aligned} & \min && k \\ & P', k : && P' \leq kQ \\ & && |P' - P| \leq e \\ & && \text{Tr}(P') = 1 \\ & && P' \geq 0 \end{aligned}$$

may be rewritten as

$$\begin{aligned} & \min && \min && k \\ & P' : && k : && P' \leq kQ \\ & && && |P' - P| \leq e \\ & && && \text{Tr}(P') = 1 \\ & && && P' \geq 0 \end{aligned}$$

By the previous lemma this is equal to

$$\begin{aligned} & \min && \max && \text{Tr}(MP') \\ & P' : && M : && \text{Tr}(MQ) = 1 \\ & && && |P' - P| \leq e \\ & && && \text{Tr}(P') = 1 \\ & && && P' \geq 0 \\ & && && M \geq 0 \end{aligned}$$

Min-max duality

A powerful min-max theorem from game theory implies

$$\begin{array}{ll}
 \min & \max \\
 P' : |P' - P| \leq e & M : \text{Tr}(MQ) = 1 \\
 \text{Tr}(P') = 1 & M \geq 0 \\
 P' \geq 0 &
 \end{array}
 \text{Tr}(MP')$$

is equal to

$$\begin{array}{ll}
 \max & \min \\
 M : \text{Tr}(MQ) = 1 & P' : |P' - P| \leq e \\
 M \geq 0 & \text{Tr}(P') = 1 \\
 & P' \geq 0
 \end{array}
 \text{Tr}(MP')$$

To bound the optimum, it suffices to produce a suitable P' for each given M with bounded $\text{Tr}(MP')$

The bound we seek: $\text{Tr}(MP') \leq 2^{(s+1)/e} / (1-e)$, where $s = S(P|Q)$

Lemma: For any $M \geq 0$ such that $\text{Tr}(MQ) = 1$, there is a quantum state P' ϵ -close to P such that

$$\text{Tr}(MP') \leq 2^{(s+1)/\epsilon} / (1-\epsilon),$$

where $s = S(P|Q)$.

Proof: Let $M = \sum_i m_i v_i v_i^*$ (spectral decomposition)

$$\text{Tr}(MQ) = \sum_i m_i v_i^* Q v_i \quad q_i \quad Q_I = (q_i)$$

$$\text{Tr}(MP) = \sum_i m_i v_i^* P v_i \quad p_i \quad P_I = (p_i)$$

Monotonicity of relative entropy implies

$$s_I = S(P_I|Q_I) \leq S(P|Q)$$

We apply the result for distributions to P_I, Q_I :

$$\text{Let } B = \{ i : \log_2 (p_i/q_i) \geq (s_I+1)/e \}$$

$$\text{so that } \sum_{i \in B} p_i \leq e$$

Let Π be the orthogonal projection onto $\text{Span} \{ v_i : i \notin B \}$

$$\text{So } \Pi = \sum_{i \notin B} v_i v_i^*$$

$$\text{and } \text{Tr}(\Pi P) = \sum_{i \notin B} v_i^* P v_i = \sum_{i \notin B} p_i \geq 1 - e$$

Define $P' = \Pi P \Pi / \text{Tr}(\Pi P)$

(we restrict P to the subspace with bounded p_i)

Lemma: $|P' - P| \leq \sqrt{e}$

(similar to the “gentle measurement lemma due to Winter)

Recall $B = \{ i : \log_2 (p_i/q_i) \geq (s+1)/e \}$

We have

$$\begin{aligned} \text{Tr}(MP') &= \sum_{i \notin B} m_i v_i^* P v_i / \text{Tr}(\Pi P) \\ &\leq (1/(1-e)) \sum_{i \notin B} m_i p_i \\ &\leq (1/(1-e)) \sum_{i \notin B} m_i 2^{(s+1)/e} q_i \\ &\leq [2^{(s+1)/e} / (1-e)] \sum_{i \notin B} m_i q_i \leq \text{Tr}(MQ) \\ &\leq 2^{(s+1)/e} / (1-e) \quad \blacksquare \end{aligned}$$

Remarks

- We get a stronger statement
 - use *fidelity* as a measure of distance
 - tighter bound in terms of *observational divergence*
- Optimal relationship between observational divergence and relative min-entropy (up to a constant factor)
- An alternative proof, albeit more abstract, using semidefinite programming duality
- Generalizes to smooth conditional entropy, which plays an important role in quantum cryptography and Shannon theory