

## Lecture 6: September 24, 2024

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Quantum Relative Entropy

## 6.1 Chernoff-Stein Lemma

Let's start by proving the Chernoff-Stein Lemma from the last lecture. The setup: we have a string  $x^n$ , which was sampled either from  $p^n$  or  $q^n$ , and we want to know which. To do this, we will look at the likelihood ratio test (LRT). To perform this test, we first compute

$$W(x^n) = \log \frac{p^n(x^n)}{q^n(x^n)}. \quad (6.1)$$

Note that  $W$  is a random variable, and that

- $\mathbb{E}_{x^n \sim p^n}[W] = nD(p||q)$
- $\mathbb{E}_{x^n \sim q^n}[W] = -nD(q||p)$

Then, to make the decision, we define some value  $T$  such that if  $W \geq T$ , we guess  $p^n$ , and if  $W < T$ , we guess  $q^n$ . Let  $A = \{x^n | W(x^n) \geq T\}$  be the "acceptance region".

We're interested in asymmetric hypothesis testing: we need  $p^n(A) \geq 1 - \epsilon$  (i.e. the probability that we guess  $q$  when it was actually  $p$  should be less than  $\epsilon$ ), and then  $q^n(A) \leq e^{-nR}$  for some  $R$  (the probability that we guess  $p$  when it was actually  $q$  should grow exponentially small with  $n$ ).

To decide where to set  $T$ , observe that if you set the threshold above  $nD(p||q)$ , then (in the limit of large sample sizes), we will never guess  $p$ . On the other hand, if we set the threshold below  $-nD(q||p)$ , we will never guess  $q$ . This suggests we should set  $T$  somewhere inside this range. Since we want to minimize  $q^n(A)$ , we'll pick  $T$  to be closer to this upper bound:  $T = n(D(p||q) - \delta)$ .

We will first show that this  $T$  achieves the desired bound for  $p^n(A)$ . Consider that

$$p^n(A) = \Pr_{x^n \sim p^n} \left[ \log \frac{p^n(x^n)}{q^n(x^n)} > nD(p||q) \right] \quad (6.2)$$

$$= \Pr_{x^n \sim p^n} \left[ D(p||q) - \frac{1}{n} \sum_{i=1}^n W[x_i] < \delta \right] \quad (6.3)$$

Since  $\mathbb{E}_{x \sim p}[W[x]] = D(p||q)$ , and each of the  $x_i$  are independent and identically drawn from the source, by the law of large numbers, this quantity approaches 1 as  $n$  goes to infinity. Thus, for any  $\epsilon$  and  $\delta$  we can take  $n$  large enough that  $p^n(A) \geq 1 - \epsilon$ .

Now to show that  $q^n(A)$  is small. If  $x^n \in A$ , then  $q^n(x^n) \leq e^{-T} p^n(x^n)$ . Then,

$$q^n(A) \leq e^{-T} p^n(A) \quad (6.4)$$

$$\leq e^{-T} \quad (6.5)$$

so  $R = D(p||q) - \delta$  (for any  $\delta > 0$ ).

### 6.1.1 Multiple hypothesis testing

We briefly mention another form of hypothesis testing: multiple hypothesis testing. Here, we have  $Q \subseteq \Delta_d = \{\text{prob dists on } [d]\}$ . Now, we want to distinguish between the two cases  $x^n \sim p^n$  or  $q^n$  for some  $q \in Q$ . Intuitively, it makes sense that distinguishing  $p$  from  $Q$  should be at least as hard as distinguishing  $p$  from  $q^*$ , where  $q^*$  is the distribution in  $Q$  closest to  $p$  (see figure 6.1). It turns out that it is actually equally as hard - you can distinguish with the exponential rate

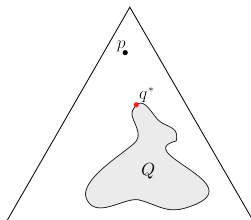


Figure 6.1: An example of multiple hypothesis testing. Given a sample  $x^n$ , we want to distinguish between two cases,  $x^n \sim p^n$ , or  $x^n \sim q^n$  where  $q \in Q$ , a subset of the probability simplex. Here,  $q^*$  is the point in  $Q$  closest to  $p$ .

$$R = \min_{q \in Q} D(p||q).$$

## 6.2 Quantum Relative Entropy and Quantum Chernoff-Stein

We will now turn to the quantum analogue of hypothesis testing. First, we define the quantum relative entropy.

**Definition 1** (Quantum Relative Entropy). *The quantum relative entropy,  $D(\rho||\sigma)$ , is defined as*

$$D(\rho||\sigma) = \text{Tr}[\rho(\log \rho - \log \sigma)].$$

Note that if  $[\rho, \sigma] = 0$ , this reduces to the classical relative entropy. Just like in the classical case,  $D(\rho||\sigma) \geq 0$ . From this, we get that

- $S(\rho) \leq d$
- $I(A; B) \geq 0$
- $S(A) \geq S(A|B)$

We also have a Quantum Pinsker's Inequality.

**Theorem 1.**

$$D(\rho||\sigma) \geq \frac{1}{2 \ln 2} \|\rho - \sigma\|_1^2.$$

Now for asymmetric hypothesis testing. Our distributions now will be two quantum states,  $\rho$  and  $\sigma$ , and the test will be a set of measurement operators  $\{M, I - M\}$  where an outcome of  $M$  means we say the state is  $\rho$ , and an outcome of  $I - M$  means we say the state is  $\sigma$ . We now want to find

$$\beta_\epsilon^n = \min \{ \text{Tr} [M \sigma^{\otimes n}] \mid \text{Tr} M \rho^{\otimes n} \geq 1 - \epsilon \}.$$

**Theorem 2** (Quantum Chernoff-Stein Theorem).

$$\lim_{n \rightarrow \infty} \frac{-1}{n} \log \beta_\epsilon^n = D(\rho||\sigma).$$

Before looking at the proof, we will examine the case when  $\rho$  and  $\sigma$  are pure and  $D(\rho||\sigma) = \infty$ , i.e.  $\text{supp}(\rho) \not\subseteq \text{supp}(\sigma)$ . Let  $\rho = |\psi\rangle\langle\psi|$  and  $\sigma = |\phi\rangle\langle\phi|$ . The measurement that achieves the desired rate is  $M = I - B^{\otimes n}$ , where  $A = |\phi^\perp\rangle\langle\phi^\perp|$ , and  $B = I - A$ . (A result of  $M$  is saying that you measured  $|\phi^\perp\rangle$  at least once).

Then

$$\text{Tr}(M\sigma^{\otimes n}) = 0$$

while  $\text{Tr}(M\rho^{\otimes n}) \rightarrow 1$  as  $n \rightarrow \infty$  (in every register you get some probability of  $|\phi^\perp\rangle$ , so as  $n \rightarrow \infty$  you are increasing your chances)

We will now prove the theorem.

*Proof.* We want an  $M$  such that  $\text{Tr}(\rho^n M) \geq \alpha$  and  $\text{Tr}(\sigma^n M) \leq e^{-nR}$ . The idea will be to construct something similar to the LRT, but we will have to be careful about eigenbases.

Let  $\rho = \sum_x r_x |\alpha_x\rangle\langle\alpha_x|$  and  $\sigma = \sum_x s_x |\beta_x\rangle\langle\beta_x|$ .

Recall the definition of a typical projector:

$$\Pi_{\rho,\delta}^n = \sum_{x^n: |\frac{1}{n} \sum_{i=1}^n \log r_{x_i} - \text{Tr}(\rho \log \rho)| \leq \delta} |\alpha_{x^n}\rangle\langle\alpha_{x^n}|.$$

Next, define

$$\Pi_{\rho||\sigma,\delta}^n = \sum_{x^n: |\frac{1}{n} \sum_{i=1}^n \log s_{x_i} - \text{Tr}(\rho \log \sigma)| \leq \delta} |\beta_{x^n}\rangle\langle\beta_{x^n}|.$$

We note that both of the subspaces defined by these projectors are typical under  $\rho$ , i.e.,  $\text{Tr}(\rho^n \Pi_{\rho,\delta}^n) \geq 1 - \epsilon$  and  $\text{Tr}(\rho^{\otimes n} \Pi_{\rho||\sigma,\delta}^n) \geq 1 - \epsilon$ . We also have that  $[\Pi_{\rho,\delta}^n, \rho^{\otimes n}] = 0$  and  $[\Pi_{\rho||\sigma,\delta}^n, \sigma^{\otimes n}] = 0$ .

If we sandwich  $\rho^{\otimes n}$  between the typical projectors, we cut off the "atypical" eigenvalues:

$$e^{-n(S(\rho)+\delta)} \Pi_{\rho,\delta}^n \leq \Pi_{\rho,\delta}^n \rho^{\otimes n} \Pi_{\rho,\delta}^n \leq e^{-n(S(\rho)-\delta)} \Pi_{\rho,\delta}^n.$$

Similarly, if you do the conditional projection, it squishes the eigenvalues of  $\sigma$  into the following range:

$$e^{n(\text{Tr}(\rho \log \sigma) - \delta)} \Pi_{\rho||\sigma} \leq \Pi_{\rho||\sigma} \sigma^n \Pi_{\rho||\sigma} \leq e^{n(\text{Tr}(\rho \log \sigma) + \delta)} \Pi_{\rho||\sigma}. \quad (6.6)$$

(Note that from here on we will drop the  $\delta$  and  $n$  on the typical projectors).

To get some intuition for equation 6.6, suppose you measure  $\log \sigma = \sum_x \log s_x \beta_x$  on  $\rho$ . Then

$$\Pr[\log s_x] = \text{Tr}[\rho \beta_x],$$

and the expectation is  $\text{Tr} \rho \log \sigma$ . If you do this  $n$  times, the law of large numbers says that the average will approach the expectation.

We will first show achievability. Our measurement will be the product of both projectors - first measure  $\{\Pi_{\rho||\sigma}, I - \Pi_{\rho||\sigma}\}$ , and if you get the positive outcome  $\Pi_{\rho||\sigma}$ , then measure  $\{\Pi_\rho, I - \Pi_\rho\}$ .

More rigorously, define

$$M = \Pi_{\rho||\sigma} \Pi_\rho \Pi_{\rho||\sigma}.$$

Then

$$\text{Tr}(\rho^{\otimes n} M) = \text{Tr}(\Pi_\rho \Pi_{\rho||\sigma} \rho^{\otimes n} \Pi_{\rho||\sigma} \Pi_\rho).$$

We need to show that the probability of  $\rho$  accepting is large. As we saw above, the probability of  $\rho$  accepting for each individual measurement is large. To show the combination works, we can

use the Gentle Measurement lemma, which says that the state after accepting  $\Pi_{\rho|\sigma}$  is still very close to  $\rho$ :

$$\|\rho^{\otimes n} - \Pi_{\rho|\sigma}\rho^{\otimes n}\Pi_{\rho|\sigma}\|_1 \leq 2\sqrt{\epsilon}.$$

We then see that

$$\text{Tr}(\Pi_{\rho}(\rho^{\otimes n} - \Pi_{\rho|\sigma}\rho^{\otimes n}\Pi_{\rho|\sigma})) \leq \frac{1}{2}\|\rho^{\otimes n} - \Pi_{\rho|\sigma}\rho^{\otimes n}\Pi_{\rho|\sigma}\|_1 \leq \sqrt{\epsilon}$$

from which it follows that

$$\text{Tr}(\Pi_{\rho}\Pi_{\rho|\sigma}\rho^{\otimes n}\Pi_{\rho|\sigma}\Pi_{\rho}) \geq \text{Tr}(\Pi_{\rho}\rho^{\otimes n}) - \sqrt{\epsilon} \geq 1 - \epsilon - \sqrt{\epsilon}.$$

So we have now showed that type one error is small enough, and now we need to bound type two error. Consider that

$$\text{Tr}(M\sigma^{\otimes n}) = \text{Tr}(\Pi_{\rho}\Pi_{\rho|\sigma}\sigma^{\otimes n}\Pi_{\rho|\sigma}) \quad (6.7)$$

$$\leq \text{Tr}\left(\Pi_{\rho}e^{n(\text{Tr}(\rho \log \sigma) + \delta)}\Pi_{\rho|\sigma}\right) \quad (6.8)$$

$$\leq e^{n(S(\rho) + \delta)}e^{n(\text{Tr}(\rho \log \sigma) + \delta)} \quad (6.9)$$

$$= e^{-n(D(\rho|\sigma) - 2\delta)} \quad (6.10)$$

where we used the operator inequality from equation 6.6, the fact that  $\Pi_{\rho|\sigma} \leq I$ , and that  $\text{Tr}(\Pi_{\rho}) = |T_{\rho}| \leq e^{n(S(\rho) + \delta)}$ .

We have now shown achievability. Next, we will use similar arguments to show the converse, i.e., you cannot do better than the rate  $D(\rho|\sigma)$ .

Suppose  $\text{Tr}(M\rho^{\otimes n}) \geq \alpha$ . Our goal is to show that  $\text{Tr}(M\sigma^{\otimes n})$  is "not too small". We will use the following operator inequalities:

$$\sigma^{\otimes n} \geq \Pi_{\rho|\sigma}e^{n(\text{Tr}(\rho \log \sigma) - \delta)} \quad (6.11)$$

$$\Pi_{\rho}\rho^{\otimes n}\Pi_{\rho} \leq e^{-n(S(\rho) - \delta)}\Pi_{\rho}. \quad (6.12)$$

Now

$$\text{Tr}(M\sigma^{\otimes n}) \geq \text{Tr}(\Pi_{\rho|\sigma}M)e^{n(\text{Tr}(\rho \log \sigma) - \delta)} \quad (6.13)$$

$$\geq (\alpha - \sqrt{2\epsilon})e^{-n(D(\rho|\sigma) - 2\delta)} \quad (6.14)$$

and

$$\text{Tr}(\Pi_{\rho|\sigma}M) \geq \text{Tr}(\Pi_{\rho|\sigma}M\Pi_{\rho|\sigma}\Pi_{\rho}) \quad (6.15)$$

$$\geq \text{Tr}\left(M\Pi_{\rho|\sigma}\rho^{\otimes n}\Pi_{\rho|\sigma}e^{-n(S(\rho) - \delta)}\right). \quad (6.16)$$

We can again use Gentle Measurement to show that  $\Pi_{\rho|\sigma}\rho^{\otimes n}\Pi_{\rho|\sigma}$  is close to  $\rho^{\otimes n}$ , and using a similar argument as before, we get that

$$\text{Tr}(M\Pi_{\rho|\sigma}\rho^{\otimes n}\Pi_{\rho|\sigma}) \geq \alpha - \sqrt{2\epsilon}.$$

Putting it all together, we find

$$\text{Tr}(M\sigma^{\otimes n}) \geq (\alpha - \sqrt{2\epsilon})e^{-n(D(\rho|\sigma) - \delta)} \quad (6.17)$$

which completes the proof.  $\square$