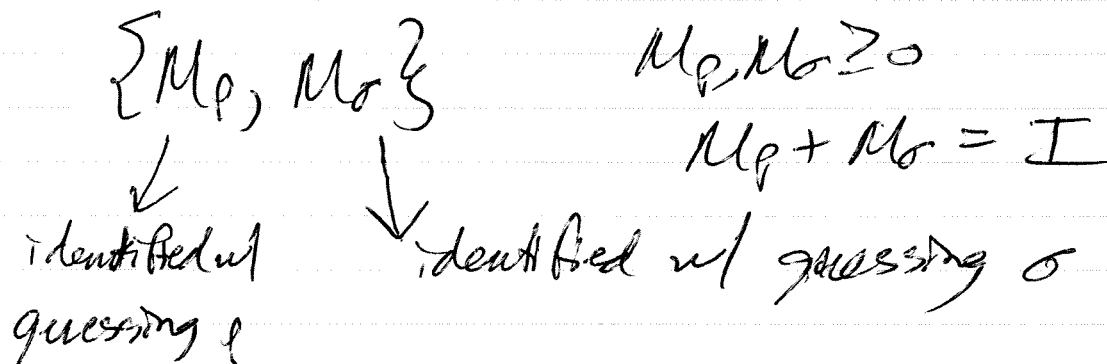


# Lecture 15

①

## Q. Hypothesis testing

q. system prepared in state  $\rho$  or  $\sigma$  & objective is to guess which one was prepared by means of a q. measurement



2 kinds of errors that can occur:

Type I: guess " $\sigma$ ", but  $\rho$  was prepared  
probability of type I error =  $\text{Tr}[M_\sigma \rho]$

Type II error: guess "p", when prepared state is  $\sigma$  (2)

$$\text{Probability of Type II error} = \text{Tr}[M_p \sigma]$$

Two different ways of combining these kinds of errors to arrive at a single notion of error:

Symmetric / Bayesian:

prior probability  $\lambda \in [0, 1]$  for being  $p$  &  $1-\lambda$  for being  $\sigma$

$$\begin{aligned} \text{Pr}[\text{error}] &= \text{Pr}[\text{error} | p] \text{Pr}[p] \\ &\quad + \text{Pr}[\text{error} | \sigma] \text{Pr}[\sigma] \\ &= \text{Tr}[(I - M) \rho] \lambda + \\ &\quad \text{Tr}[M \sigma] \cdot (1 - \lambda) \\ &\equiv P_{\text{err}}(\lambda, \rho, \sigma, M) \end{aligned}$$

goal is to minimize this error prob.

optimization problem is then

(3)

$$P_{\text{err}}^*(\lambda, \rho, \sigma) = \min_M \left\{ P_{\text{err}}(\lambda, \rho, \sigma, M) : 0 \leq M \leq I \right\}$$

This is an SDP:

~~can~~ can write

$$P_{\text{err}}(\lambda, \rho, \sigma, M) \\ = \lambda + \text{Tr}[M((1-\lambda)\sigma - \lambda\rho)]$$

$$\Rightarrow P_{\text{err}}^*(\lambda, \rho, \sigma) = \min_{M \geq 0} \left\{ \text{Tr}[M((1-\lambda)\sigma - \lambda\rho)] : M \leq I \right\} + \lambda$$

other kind of hypothesis testing setting is

Asymmetric case: minimize Type II error prob. subject to constraint on

Type I error prob.:

$$\min_M \left\{ \text{Tr}[M\sigma] : \text{Tr}[(I-M)\rho] \leq \varepsilon, 0 \leq M \leq I \right\}$$

(4)

This is also an SDP!

---

Let us focus on the symmetric case 1st.

property of  $P_{err}^*$ :

obeys data processing inequality, i.e.,

$$P_{err}^*(\rho, \sigma) \leq P_{err}^*(\rho, \mathcal{N}(\rho), \mathcal{N}(\sigma))$$

Intuition: if you act w/ a channel 1st before doing an optimal measurement, then error cannot decrease.

Proof: let  $\mathcal{M}'$  be an arbitrary measurement op. Then  $\mathcal{N}^+(\mathcal{M}')$  is also a measurement op.

5

$$\begin{aligned} \Rightarrow & \lambda + \text{Tr}[M'((1-\lambda)\sigma) - \lambda N(\rho)] \\ &= \lambda + \text{Tr}[N+(M')((1-\lambda)\sigma) - \lambda\rho] \\ &\geq \lambda + \min_{0 \leq M \leq I} \text{Tr}[M((1-\lambda)\sigma) - \lambda\rho] \\ &= P_{\text{err}}^*(\lambda, \rho, \sigma) \end{aligned}$$

Since ineq. holds for all measurement op's, conclude that

$$\begin{aligned} P_{\text{err}}^*(\lambda, N(\rho), N(\sigma)) \\ \geq P_{\text{err}}^*(\lambda, \rho, \sigma) \end{aligned}$$

closed-form expression for

$P_{\text{err}}^*(\lambda, \rho, \sigma)$  via

Helstrom - Holevo theorem:

$$\inf_M \left\{ \text{Tr}[(I-M)A] + \text{Tr}[MB] : 0 \leq M \leq I \right\} \\ = \frac{1}{2} (\text{Tr}[A+B] - \|A-B\|_1)$$

(6)

Let  $M$  be an arbitrary measurement  $\rho$ .

Set  $\Delta = A - B$  & let  $\Delta_+$  &

$\Delta_-$  be positive & negative parts

of  $\Delta$ . ( $A - B = \Delta_+ - \Delta_-$  &

$$\Delta_+ \Delta_- = 0)$$

can write

$$\text{Tr}[(I-M)A] + \text{Tr}[MB]$$

$$= \text{Tr}[A] - (\text{Tr}[M\Delta_+] - \text{Tr}[M\Delta_-])$$

since  $\text{Tr}[M\Delta_-] \geq 0$ ,

follows that

$$\text{Tr}[M\Delta_+] - \text{Tr}[M\Delta_-] \leq \text{Tr}[M\Delta_+]$$

$$\Rightarrow \text{Tr}[M\Delta_+] - \text{Tr}[M\Delta_-] \leq \text{Tr}[\Delta_+]$$

$$\text{Tr}[(I-M)A] + \text{Tr}[MB] \geq \text{Tr}[A] - \text{Tr}[\Delta_+]$$

$$\text{Since } \|A - B\|_1 = \text{Tr}[|A - B|]$$

$$= \text{Tr}[\Delta_+] + \text{Tr}[\Delta_-]$$

$$\Delta_- = \Delta_+ + B - A$$

$$\Rightarrow \text{Tr}[\Delta_+] = \frac{1}{2} (\|A-B\|_1 - \text{Tr}[B-A]) \quad (17)$$

$$\begin{aligned} &\Rightarrow \\ &\inf \text{Tr}[(I-M)A] + \text{Tr}[MB] \\ &M: 0 \leq M \leq I \geq \frac{1}{2} (\text{Tr}[A+B] - \|A-B\|_1) \end{aligned}$$

---

To see reverse inequality,

pick  $M = \Pi_+$  (projection onto positive part of  $A-B$ )

$$\text{Then } \text{Tr}[M\Delta_+] = \text{Tr}[\Delta_+] +$$

$$\text{Tr}[M\Delta_-] = 0$$

Then we get equality

8

Pick  $A = \frac{1}{2} \rho$      $B = \frac{1}{2} \sigma$

$\Rightarrow \text{Perr}^x(\frac{1}{2}, \rho, \sigma)$

$= \frac{1}{2} (1 - \frac{1}{2} \|\rho - \sigma\|_1)$

↑  
gives operational meaning  
to trace distance.