

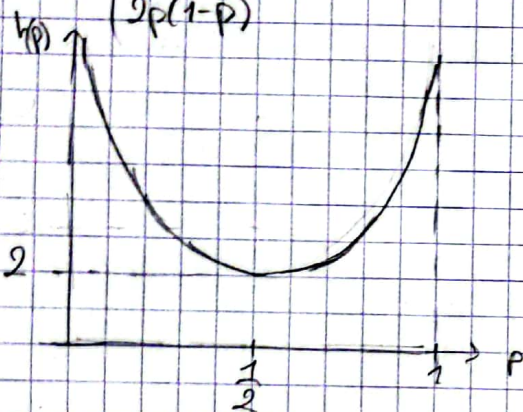
# Detection Theory - Homework 2

Nicola's Violante

•  $X_1, \dots, X_\ell$  independent random variables,  $0 \leq X_i \leq 1$ , i.i.d

•  $S_\ell := \sum_{i=1}^{\ell} X_i$ ,  $p := E\left[\frac{S_\ell}{\ell}\right]$

1) 
$$h(p) = \begin{cases} \frac{1}{1-2p} \log\left(\frac{1-p}{p}\right) & \text{if } 0 < p < \frac{1}{2} \\ \frac{1}{2p(1-p)} & \text{if } \frac{1}{2} \leq p < 1 \end{cases}$$



2)  $X$  r.v. /  $a \leq X \leq b$ ,  $\lambda > 0$

•  $\frac{b-x}{b-a} + \frac{x-a}{b-a} = 1$

• Since  $a \leq X \leq b \Rightarrow \frac{b-x}{b-a} \geq 0$  and  $\frac{x-a}{b-a} \geq 0$

$$e^{\lambda \left( \frac{b-x}{b-a} \cdot a + \frac{x-a}{b-a} \cdot b \right)} \leq \frac{b-x}{b-a} \cdot e^{\lambda a} + \frac{x-a}{b-a} \cdot e^{\lambda b} \quad \text{by convexity}$$

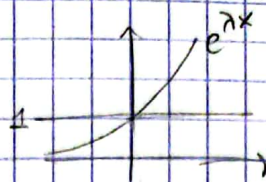
$$\Rightarrow e^{\lambda x} \leq \left( \frac{b-x}{b-a} \right) e^{\lambda a} + \left( \frac{x-a}{b-a} \right) e^{\lambda b}$$

$$\Rightarrow E(e^{\lambda x}) \leq \left( \frac{b - E(x)}{b-a} \right) e^{\lambda a} + \left( \frac{E(x) - a}{b-a} \right) e^{\lambda b}$$



3) a)  $\mathbb{1}_{x \geq 0} \leq e^{\lambda x}$

- At  $x=0$ :  $e^{\lambda \cdot 0} = 1$
- if  $\lambda > 0$  and  $x > 0$  then  $\lambda x > 0$  which implies that  $e^{\lambda x} > 1$



b)  $\mathbb{1}_{\{S_e - \mathbb{E}(S_e) - \ell t \geq 0\}} \geq e^{\lambda(S_e - \mathbb{E}(S_e) - \ell t)}$

Taking expectation

$$\mathbb{E}(\mathbb{1}_{\{S_e - \mathbb{E}(S_e) - \ell t \geq 0\}}) \geq \mathbb{E}(e^{\lambda(S_e - \mathbb{E}(S_e) - \ell t)})$$

$$= \underbrace{e^{-\lambda(\mathbb{E}(S_e) + \ell t)}}_{= e^{-\lambda \ell p}} \mathbb{E}(e^{\lambda S_e})$$

$$= \mathbb{P}(S_e - \mathbb{E}(S_e) - \ell t \geq 0)$$

$$\Rightarrow \mathbb{P}(S_e \geq \ell(p+t)) \geq e^{-\lambda \ell(p+t)} \mathbb{E}(e^{\lambda \sum_{i=1}^{\ell} X_i})$$

$$= e^{-\lambda \ell(p+t)} \mathbb{E}\left(\prod_{i=1}^{\ell} e^{\lambda X_i}\right)$$

$$= e^{-\lambda \ell(p+t)} \prod_{i=1}^{\ell} \mathbb{E}(e^{\lambda X_i}) \quad \left\{ \begin{array}{l} X_i \text{ are independent} \end{array} \right.$$

4)  $p_i = \mathbb{E}(X_i)$

a)  $\frac{\ell}{\prod_{i=1}^{\ell}} \mathbb{E}(e^{\lambda X_i}) \leq \frac{\ell}{\prod_{i=1}^{\ell}} (1 - \mathbb{E}(X_i) + \mathbb{E}(X_i) e^{\lambda})$  by part b)

$$= \frac{\ell}{\prod_{i=1}^{\ell}} (1 - p_i + p_i e^{\lambda})$$

b) If  $X_i$  are Bernoulli then  $X_i \in \{0, 1\}$  and the equality holds



$$5) \left( \prod_{i=1}^l a_i \right)^{1/l} \leq \frac{1}{l} \sum_{i=1}^l a_i \quad a_i \geq 0$$

$$\bullet \frac{1}{l} \geq 0 \text{ and } \sum_{i=1}^l \frac{1}{l} = 1$$

$$\left. \begin{aligned} &\text{By Jensen's inequality } \log \sum_{i=1}^l \frac{1}{l} a_i \geq \sum_{i=1}^l \frac{1}{l} \log a_i \\ &\log \left( \frac{1}{l} \prod_{i=1}^l a_i \right)^{1/l} = \frac{1}{l} \sum_{i=1}^l \log a_i \end{aligned} \right\} \Rightarrow$$

$$\Rightarrow \log \sum_{i=1}^l \frac{1}{l} a_i \geq \log \left( \frac{1}{l} \prod_{i=1}^l a_i \right)^{1/l} \Rightarrow \sum_{i=1}^l \frac{1}{l} a_i \geq \left( \frac{1}{l} \prod_{i=1}^l a_i \right)^{1/l}$$

$$\begin{aligned} 6) \frac{1}{l} \sum_{i=1}^l \mathbb{E}(e^{\lambda x_i}) &\leq \left( \frac{1}{l} \sum_{i=1}^l (1 - p_i + p_i e^{\lambda}) \right)^{1/l} \\ &\leq \left( \frac{1}{l} \sum_{i=1}^l (1 - p + p e^{\lambda}) \right)^{1/l} \\ &= \left( 1 - \sum_{i=1}^l \frac{p_i}{l} + e^{\lambda} \sum_{i=1}^l \frac{p_i}{l} \right)^{1/l} \\ &= \left( 1 - \mathbb{E}\left(\frac{S_l}{l}\right) + e^{\lambda} \mathbb{E}\left(\frac{S_l}{l}\right) \right)^{1/l} = (1 - p + p e^{\lambda})^{1/l} \end{aligned}$$

7) Combining 3) and 6)

$$P(S_l \geq (p+t)l) \leq e^{-\lambda(p+t)l} (1 - p + p e^{\lambda})^l = [e^{-\lambda(p+t)} (1 - p + p e^{\lambda})]^l$$

Taking  $\log$  on the right side:  $l [-\lambda(p+t) + \log(1 - p + p e^{\lambda})] =: f(\lambda)$

$$\Rightarrow \text{we derivate } \frac{\partial f}{\partial \lambda} = l \left[ -(p+t) + \frac{p e^{\lambda}}{1 - p + p e^{\lambda}} \right] = 0$$

$$\Leftrightarrow (p+t)(1 - p + p e^{\lambda}) = p e^{\lambda}$$

$$\text{if } \lambda = \log \frac{(1-p)(p+t)}{(1-p-t)p} \quad \bullet \quad p e^{\lambda} = \frac{p(1-p)(p+t)}{(1-p-t)p} = \frac{(1-p)(p+t)}{1-p-t}$$

$$\bullet (p+t)(1 - p + p e^{\lambda}) = (p+t)(1-p) \left( 1 + \frac{(p+t)}{1-p-t} \right) = \frac{(p+t)(1-p)}{1-p-t}$$

$\Rightarrow$  the upper-bound is minimized for this value of  $\lambda$ .



2)  $\lambda = \log \frac{(1-p)(p+t)}{(1-p-t)p}$  for  $0 < t < 1-p$

$\lambda > 0 \Leftrightarrow (1-p)(p+t) > (1-p-t)p \Leftrightarrow p+t-p^2-p^2 > p-p^2-\frac{1}{p}$   
 $\Leftrightarrow t > 0 \checkmark$

8)  $G(t, p) = \frac{p+t}{t^2} \log \frac{(p+t)}{p} + \frac{1-p-t}{t^2} \log \frac{(1-p-t)}{1-p}$

•  $t^2 \frac{\partial G}{\partial t} = t^2 \left\{ \frac{(t^2 - (p+t)2t)}{t^4} \log \frac{(p+t)}{p} + \frac{(p+t)}{t^2} \frac{1}{p} - \frac{1}{p} \right.$   
 $\left. + \frac{-t^2 - (1-p-t)2t}{t^4} \log \frac{(1-p-t)}{1-p} - \frac{(1-p-t)}{t^2} \frac{1}{1-p} \cdot \frac{1}{1-p} \right\}$   
 $= \left[ 1 - 2 \frac{(p+t)}{t} \right] \log \left( \frac{p+t}{p} \right) - \left[ 1 + 2 \frac{(1-p-t)}{t} \right] \log \left( \frac{1-p-t}{1-p} \right)$   
 $= - \left[ 1 - 2 \frac{(p+t)}{t} \right] \log \left( 1 - \frac{t}{t+p} \right) - \left[ -1 + 2 \frac{(1-p)}{t} \right] \log \left( 1 - \frac{t}{1-p} \right)$   
 $= \left[ 1 - 2 \frac{(1-p)}{t} \right] \log \left( 1 - \frac{t}{1-p} \right) - \left[ 1 - 2 \frac{(p+t)}{t} \right] \log \left( 1 - \frac{t}{t+p} \right)$   
 $= H \left( \frac{t}{1-p} \right) - H \left( \frac{t}{t+p} \right)$

•  $H(x) = \left( 1 - \frac{2}{x} \right) \log(1-x)$

The McLaurin expansion of  $\log(1-x)$  is  $-x - \frac{x^2}{2} - \frac{x^3}{3} - \dots$

$H(x) = \left( 1 - \frac{2}{x} \right) \sum_{i=1}^{\infty} \frac{-x^i}{i} = \sum_{i=1}^{\infty} \frac{-x^i}{i} + 2 \sum_{i=1}^{\infty} \frac{x^{i-1}}{i} = 2 + \sum_{i=2}^{\infty} \left( \frac{2}{i+1} - \frac{1}{i} \right) x^i$

• For  $0 < x < 1$ : the terms  $\left( \frac{2}{i+1} - \frac{1}{i} \right) > 0$  and so  $H'(x) > 0 \Rightarrow$  it's increasing

•  $\frac{\partial G(t, p)}{\partial t} = \frac{1}{t^2} \left[ H \left( \frac{t}{1-p} \right) - H \left( \frac{t}{t+p} \right) \right] > 0 \Leftrightarrow \frac{t}{1-p} > \frac{t}{t+p}$  since  $H$  increases  
 $\Leftrightarrow t+p > 1-p$   
 $\Leftrightarrow t > 1-2p$



# Detection Theory - Homework 2

Nirad's Violante

8) When  $t = 1 - 2p \Rightarrow \frac{\partial G}{\partial t} = 0$

$$\begin{aligned} G(1-2p, p) &= \frac{p + 1 - 2p}{(1-2p)^2} \log\left(\frac{p + 1 - 2p}{p}\right) + \frac{(1-p) - 1 + 2p}{(1-2p)^2} \log\left(\frac{1-p - 1 + 2p}{1-p}\right) \\ &= \frac{(1-p)}{(1-2p)^2} \log\left(\frac{1-p}{p}\right) + \frac{p}{(1-2p)^2} \log\left(\frac{p}{1-p}\right) \\ &= \frac{(1-p-p)}{(1-2p)^2} \log\left(\frac{1-p}{p}\right) = \frac{1}{1-2p} \log\left(\frac{1-p}{p}\right) \end{aligned}$$

9)  $1 - 2p \leq 0$

Since  $G$  is increasing (on  $t$ ) the minimum is attained when  $t \rightarrow 0$