

## Various feedback on the course:

By Ludovic Schwartz:

a convex function  $f: X \rightarrow \mathbb{R}$  (or even  $[0,1]$ ) does not necessarily achieve its infimum,

hence we safely and rightfully considered  $p_1^*$  as an approximate arginf in the last proof of Lecture 1

Ex:  $f(x) = x^2 + \frac{1}{q} \mathbb{1}_{x=0}$ , infimum equal to 0, never achieved  
(take  $f(x) = x^2 + \frac{1}{q} \mathbb{1}_{x=0}$  for an exp-concave example)

By Hassan Striber:

in the last proof it is easy to directly optimize over  $\varepsilon \in (0,1)$  the obtained bound:

$$B(\varepsilon) = T \ln(1-\varepsilon) + (N-1) \ln \varepsilon$$

$$B'(\varepsilon) = \frac{-T}{1-\varepsilon} + \frac{N-1}{\varepsilon}$$

$$B''(\varepsilon) < 0$$

thus the  $\varepsilon^*$  s.t.  $B'(\varepsilon^*) = 0$  is the global minimum of  $B$

$$\text{We have } B'(\varepsilon^*) = 0 \Leftrightarrow \frac{\varepsilon^*}{N-1} = \frac{1}{T} - \frac{\varepsilon^*}{T} \Leftrightarrow \varepsilon^* = \frac{1/T}{1/T + 1/(N-1)} = \frac{N-1}{T+N-1}$$

We thus get the sharper final bound

$$\begin{aligned} & \frac{1}{\eta} \left( T \ln \left( \frac{T+N-1}{T} \right) + (N-1) \ln \left( \frac{T+N-1}{N-1} \right) \right) = \frac{T+N-1}{\eta} H \left( \frac{N-1}{T+N-1} \right) \\ & = \frac{1}{\eta} \left( (N-1) \ln \left( 1 + \frac{N-1}{T} \right) + (N-1) \ln \left( 1 + \frac{T}{N-1} \right) \right) \end{aligned}$$

where  $H(x) = -x \ln x - (1-x) \ln(1-x)$  is the binary entropy

$\leq \frac{(N-1)^2}{T}$

\* However \* both bounds are  $\sim \frac{N-1}{\eta} \ln T$  as  $T \rightarrow +\infty$ .

Proposition de meilleure solution la question 2 de l'exercice 3

Auteur : Hassan Saber (non relu en détails par Gilles Stoltz)

\* Calcul de  $\inf_{\varepsilon \in (0,1)} -\ln(\varepsilon^\alpha e^{-\beta\varepsilon})$  pour  $\alpha, \beta > 0$

$$\text{Avec } f(\varepsilon) = -\ln(\varepsilon^\alpha e^{-\beta\varepsilon}) \text{ pour } \varepsilon \in (0,1) \\ = \alpha(-\ln \varepsilon) + \beta\varepsilon$$

$$f(0) = +\infty, \quad f(1) = \beta$$

$$\text{Ainsi } \left\{ \begin{array}{l} \text{ou bien } \arg\min_{\varepsilon \in (0,1)} f(\varepsilon) = 1 \text{ et } \inf_{\varepsilon \in (0,1)} f(\varepsilon) = \beta \\ \text{ou bien } \varepsilon^* = \arg\min_{\varepsilon \in (0,1)} f(\varepsilon) \text{ et } f'(\varepsilon^*) = 0 \end{array} \right.$$

$$f'(\varepsilon^*) = -\frac{\alpha}{\varepsilon^*} + \beta = 0 \Leftrightarrow \varepsilon^* = \frac{\alpha}{\beta} \quad (1)$$

$$(1) \text{ n'est possible que si } \frac{\alpha}{\beta} < 1 \text{ alors } f(\varepsilon^*) = \alpha \left( \ln\left(\frac{\beta}{\alpha}\right) + 1 \right) \\ \leq \alpha \left( \frac{\beta}{\alpha} - 1 + 1 \right) \\ \leq \beta$$

Ainsi, on a montré que

$$\inf_{\varepsilon \in (0,1)} f(\varepsilon) = \begin{cases} \alpha \left( \ln\left(\frac{\beta}{\alpha}\right) + 1 \right) & \text{si } \frac{\beta}{\alpha} > 1 \\ \beta & \text{si } \frac{\beta}{\alpha} \leq 1 \end{cases}$$

On cherche un encadrement de  $\inf_{\eta > 0} Q(\eta)$

où  $Q(\eta) = \frac{\eta}{8} (H-m)^2 T + \frac{1}{8} \inf_{\varepsilon \in (0,1)} \left\{ -\ln \varepsilon^{N-2} e^{-\eta T (H-m)} \varepsilon \right\}$

On pose  $\alpha = N-2$ ,  $\beta = \eta T (H-m)$  et  $\gamma = \frac{T(H-m)}{N-2}$

D'après ce qui précède, si  $\underline{\eta\gamma \leq 1}$ ,  $Q(\eta) = \frac{\eta}{8} (H-m)^2 T (H-m)$

d'où  $\inf_{\eta\gamma \leq 1} Q(\eta) = T(H-m)$

• si  $\underline{\eta\gamma > 1}$ ,  $Q(\eta) = (\eta\gamma) \frac{(H-m)(N-2)}{8} + \frac{1}{(\eta\gamma)} T(H-m) (\ln(\eta\gamma) + 1)$

d'où  $\inf_{\eta\gamma > 1} Q(\eta) = \inf_{\eta\gamma > 1} \eta \frac{(H-m)(N-2)}{8} + \frac{T(H-m)}{\eta} (\ln(\eta\gamma) + 1)$

$$= \frac{(H-m)(N-2)}{4} \inf_{\eta\gamma > 1} \frac{1}{2} \eta + \frac{a^2}{\eta} (\ln(\eta) + 1)$$

avec  $a^2 = \frac{4T}{N-2}$

On pose alors  $g(\eta) = \frac{1}{2} \eta + \frac{a^2}{\eta} (\ln(\eta) + 1)$

On encadre alors  $\inf_{\eta > 0} g(\eta)$

g est  $\mathcal{C}^2$  et  $g'(y) = \frac{1}{2} - \frac{a^2}{y^2} (\ln(y+2)) + \frac{a^2}{y^2}$

$$= \frac{\frac{1}{2} y^2 - a^2 \ln(y)}{y^2}$$

On pose  $h(y) = \frac{1}{2} y^2 - a^2 \ln(y)$  pour  $y \geq 1$ .

$h$  est  $\mathcal{C}^2$  et  $h'(y) = y - \frac{a^2}{y}$

$$h''(y) = 1 + \frac{a^2}{y^2} \geq 0 \quad h'(a) = 0$$

D'où

	1	a	$+\infty$
$h''$		+	+
$h'$	-	0	+
$h$	$1/2$	$h(a)$	$+\infty$

• Si  $h(a) = \frac{1}{2} a^2 (1 - \ln(a^2)) \geq 0 \Leftrightarrow a^2 \leq \exp(1)$

alors  $h \geq 0$  et  $g' \geq 0$  et  $\inf_{y \geq 1} g(y) = g(1) = \frac{1}{2} + a^2$



Base inférieure  $g(y_0)$

$$\frac{1}{2} y_0^2 = a^2 \ln(y_0) \Rightarrow y_0 = a \sqrt{2 \ln(y_0)}$$

$$\text{d'où } \frac{1}{2} y_0^2 = a^2 \ln(y_0) = a^2 \ln(a) + \frac{a^2}{2} \ln(2) + \frac{1}{2} a^2 \ln \ln(y_0)$$

$$\text{Or } y_0 > a \text{ et } a > \exp(2) \text{ d'où } \ln \ln(y_0) > 0$$

$$\text{et } \frac{1}{2} y_0^2 \geq a^2 \ln(a) + \frac{\ln(2)}{2}$$

$$\text{i.e. } y_0 \geq y_1 := a \sqrt{2 \ln(a) + \ln(2)}$$

$$\text{Or } a \leq y_1 \leq y_0 \text{ donc } \psi(a) \leq \psi(y_1) \leq \psi(y_0) = g(y_0)$$

$$\text{d'où } g(y_0) \geq a \sqrt{2 \ln(a) + \ln(2)} + \frac{a}{\sqrt{2 \ln(a) + \ln(2)}}$$

Base supérieure sur  $g(y_0)$ .  $\varepsilon > 0$

$$\begin{aligned} h(a^{1+\varepsilon}) &= \frac{1}{2} a^{2+2\varepsilon} - a^2 \ln(a^{1+\varepsilon}) \\ &= \frac{1}{2} a^2 \left[ (a^2)^\varepsilon - (1+\varepsilon) \ln(a^2) \right] \\ &= \frac{1}{2} a^2 \left[ e^{\varepsilon \ln(a^2)} - (1+\varepsilon) \ln(a^2) \right] \\ &\geq \frac{1}{2} a^2 \left[ 1 + \varepsilon \ln(a^2) + \frac{\varepsilon^2 (\ln(a^2))^2}{2} - (1+\varepsilon) \ln(a^2) \right] \end{aligned}$$

$$\begin{aligned}
 d'au \quad h(a^{1+\varepsilon}) &\geq \frac{1}{2} a^2 \left[ 1 + \ln(a^2) \left( \frac{\varepsilon^2 \ln(a^2)}{2} - 1 \right) \right] \\
 &\geq \frac{1}{2} a^2 \ln(a^2) \left( \frac{\varepsilon^2 \ln(a^2)}{2} - 1 \right)
 \end{aligned}$$

$$\text{Ainsi} \quad \frac{\varepsilon^2 \ln(a^2)}{2} = 1, \quad \varepsilon = \frac{1}{\sqrt{\ln(a)}} \quad \text{et} \quad h(a^{1+\varepsilon}) \geq 0$$

$$\text{Comme} \quad a < a^{1+\varepsilon} \quad \text{et} \quad h(a^{1+\varepsilon}) \geq 0, \quad \text{on a} \quad y_0 \leq y_\varepsilon = a^{1+\frac{1}{\sqrt{\ln(a)}}}$$

$$\text{Or} \quad \frac{1}{2} y_0^2 = a^2 \ln(a) + \frac{a^2}{2} \ln(2) + \frac{1}{2} a^2 \ln \ln(y_0)$$

$$\begin{aligned}
 d'au \quad \frac{1}{2} y_0^2 &\leq a^2 \ln(a) + \frac{a^2}{2} \ln(2) + \frac{1}{2} a^2 \ln \ln(y_\varepsilon) \\
 &= \ln \left( \left( 1 + \frac{1}{\sqrt{\ln(a)}} \right) \ln(a) \right) \\
 &= \ln \left( \ln(a) + \sqrt{\ln(a)} \right) \\
 &\leq \ln(2 \ln(a))
 \end{aligned}$$

$$\text{Ainsi} \quad y_0 \leq y_3 := a \sqrt{2 \ln(a) + \ln(2 \ln(a)) + \ln(2)}$$

$$\text{Et} \quad g(y_0) = \varphi(y_0) \leq \varphi(y_3) = a \sqrt{2 \ln(a) + \ln(2 \ln(a)) + \ln(2)} + \frac{a}{\sqrt{2 \ln(a) + \ln(2 \ln(a)) + \ln(2)}}$$

Alors

$$\min\left(\frac{1}{2} + a^2, a \Psi\left(\sqrt{\ln(2a^2)}\right)\right) \leq \inf_{\gamma > 1} g(\gamma) \leq a \Psi\left(\sqrt{\ln(2a^2) + \ln(\ln(a^2))}\right)$$

où  $\Psi: x \mapsto x + \frac{1}{x}$ .

En remplaçant  $a$  par  $\sqrt{\frac{4T}{N-1}}$ , il vient :

$$\min\left(\frac{(H-m)(N-1)}{8} + (H-m)T; \frac{(H-m)}{2} \sqrt{T(N-1)} \Psi\left(\sqrt{\ln\left(\frac{8T}{N-1}\right)}\right)\right) \leq \inf_{\gamma > 1} Q(\gamma)$$

et  $\inf_{\gamma > 1} Q(\gamma) \leq \frac{(H-m)}{2} \sqrt{T(N-1)} \Psi\left(\sqrt{\ln\left(\frac{8T}{N-1}\right) + \ln\ln\left(\frac{4T}{N-1}\right)}\right)$

En conclusion, on a donc :

$$\begin{cases} \min\left(T(H-m); (H-m)\sqrt{T(N-1)} \Psi\left(\sqrt{\ln\left(\frac{8T}{N-1}\right)}\right)\right) \leq \inf_{\gamma > 0} Q(\gamma) \\ \inf_{\gamma > 0} Q(\gamma) \leq \frac{(H-m)}{2} \sqrt{T(N-1)} \Psi\left(\sqrt{\ln\left(\frac{8T}{N-1}\right) + \ln\ln\left(\frac{4T}{N-1}\right)}\right) \end{cases}$$

On a alors  $\inf_{\gamma > 0} Q(\gamma) \sim \frac{(H-m)}{2} \sqrt{T(N-1)} \sqrt{\ln(T)}$

$T \rightarrow \infty$