

Ex 1 (10P)

1)  $\kappa = \|I\| = \|A A^{-1}\| \leq \|A\| \|A^{-1}\| = \text{Cond}(A)$  (1P)

-  $\text{Cond}(cA) = \|cA\| \cdot \|(cA)^{-1}\| = |c| \|A\| \cdot \frac{1}{|c|} \|A^{-1}\| = \|A\| \|A^{-1}\| = \text{Cond}(A)$  (2P)

-  $\text{Cond}(AB) = \|A \cdot B\| \|(A \cdot B)^{-1}\| \leq \|A\| \|A^{-1}\| \|B\| \|B^{-1}\| = \text{Cond}(A) \cdot \text{Cond}(B)$  (2P)

e)  $\|A\|_2 = \sqrt{\sigma_n}$ ,  $\|A^{-1}\|_2 = \frac{1}{\sqrt{\sigma_1}} \Rightarrow \|A\|_2 \cdot \|A^{-1}\|_2 = \sqrt{\frac{\sigma_n}{\sigma_1}} = \text{Cond}_2(A)$  (3P)

-  $A = A^t \Rightarrow \|A\|_2 = \lambda_n$ ,  $\|A^{-1}\|_2 = \frac{1}{\lambda_1} \Rightarrow \text{Cond}_2(A) = \frac{\lambda_n}{\lambda_1}$  (2P)

-  $A = \alpha Q \Rightarrow \text{Cond}_2(A) = \text{Cond}_2(Q)$  et  $\|Q\|_2 = \|Q^{-1}\|_2 = 1$  (2P)  
~~et donc~~ donc  $\text{Cond}_2(A) = 1$

Ex 2 (10P)

1.  $J = \begin{bmatrix} 0 & -\alpha & \beta \\ 0 & 0 & 0 \\ \beta & -\alpha & 0 \end{bmatrix}$   $\det(J - \lambda I) = \lambda(\lambda^2 - \beta^2) \Rightarrow \lambda_1 = 0, \lambda_2 = +\beta, \lambda_3 = -\beta$  et  $\rho(J) = |\beta|$ .

$\rho(J) < 1 \Leftrightarrow |\beta| < 1$  et  $\alpha \in \mathbb{R}$ . (2P)

e) -  $\alpha = \beta = 2 \Rightarrow \rho(J) > 1$  Jacobson diverge. (1P)

-  $x^0 = \begin{pmatrix} 2 \\ 2 \\ 2 \end{pmatrix}$ ,  $x^1 = Jx^0 + 0^1 b = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ ,  $x = Jx + 0^1 b = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ . (2P)

Par récurrence  $x^k = \begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix} \forall k \geq 1$  car  $x^1$  est solution

de  $Ax = b \Leftrightarrow x = Jx + 0^1 b$ .

3)  $\begin{cases} x_1^{k+1} = -\alpha x_2^k + \beta x_3^k + 1 \\ x_2^{k+1} = 1 \\ x_3^{k+1} = \beta x_1^k - \alpha x_2^k + 1 = -\alpha \beta x_2^k + \beta x_3^k + \beta - \alpha + 1 \end{cases} \rightarrow (2P)$

$\Rightarrow G = \begin{bmatrix} 0 & -\alpha & \beta \\ 0 & 0 & 0 \\ 0 & -\alpha\beta & \beta \end{bmatrix} = (G - E)F \rightarrow (1P)$

4)  $\det(G - \lambda I) = \lambda(\lambda - \beta) = 0 \Rightarrow \lambda_1 = 0, \lambda_2 = 0, \lambda_3 = \beta$  et  $\rho(G) = |\beta|$ . (2P)

La méthode de Gauss-Jordan converge  $\Leftrightarrow \rho(G) < 1 \Leftrightarrow |\beta| < 1$  et  $\alpha \in \mathbb{R}$

Solution examen Final  
optimisation  
Niveau II - PA

(8P)

EX 1)  $\nabla f(x,y) = \begin{pmatrix} 6x^2 + 6y \\ 6x - 6y \end{pmatrix} \Leftrightarrow \begin{pmatrix} 0 \\ 0 \end{pmatrix} \Leftrightarrow \begin{cases} x^2 + y = 0 \\ x = y \end{cases} \Leftrightarrow \begin{cases} x^2 + x = 0 \\ x = y \end{cases} \Leftrightarrow \begin{cases} x(x+1) = 0 \\ x = y \end{cases}$

Deux points critiques  $\begin{pmatrix} 0 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ -1 \end{pmatrix}$  (2P)

2)  $H_f(x,y) = \begin{pmatrix} 12x & 6 \\ 6 & -6 \end{pmatrix}$ ,  $H_f(0,0) = \begin{pmatrix} 0 & 6 \\ 6 & -6 \end{pmatrix}$ ,  $\det H_f(0,0) = -36 < 0$ ,  $\lambda_1, \lambda_2$

de signes opposés  $\Leftrightarrow$   $f$  n'a pas de minimum local  
donc  $(0,0)$  point stationnaire (point-selle) (2P)

$H_f(-1,-1) = \begin{pmatrix} -12 & 6 \\ 6 & -6 \end{pmatrix}$ ,  $\det H_f(-1,-1) = 36 > 0$  et  $\text{Tr}(H_f(-1,-1)) = -18 < 0 \Rightarrow \lambda_1, \lambda_2 < 0$  (2P)

$H_f(-1,-1)$  défini négative  $\Leftrightarrow$   $f$  concave au voisinage  $(-1,-1) \Rightarrow (-1,-1)$  maximum local

3) -  $f$  n'a pas de minimum local  $\Rightarrow$   $f$  n'a pas de minimum global (1P)

-  $f(x,0) = 2x^3 + 2$ , donc  $f(x,0) \rightarrow +\infty$   $\Rightarrow$   $f$  n'a pas de maximum global (1P)

Problème (12P)

(1P) a)  $A = 2I$ ,  $b = \begin{pmatrix} 14 \\ 6 \end{pmatrix}$ ,  $\alpha = -7$ ,  $H(b) = 2I$  défini positive  $\Rightarrow$  strictement convexe  
(2P) b)  $(0,0) \in \Omega$ ,  $\Omega = \Theta^{-1}([-\alpha, \alpha] \times [0, \alpha])$  /  $\Theta: \begin{matrix} \mathbb{R}^2 \rightarrow \mathbb{R}^2 \\ (x_1, x_2) \rightarrow \Theta(x_1, x_2) = \begin{pmatrix} x_1 + x_2 - 2 \\ x_1 + 2x_2 - 3 \end{pmatrix} \end{matrix}$  continue

$\Omega = \{(x_1, x_2) \in \mathbb{R}^2 \mid \begin{matrix} \Theta_1(x_1, x_2) = x_1 + x_2 - 2 \leq 0 \\ \Theta_2(x_1, x_2) = x_1 + 2x_2 - 3 \leq 0 \end{matrix} \}$  /  $\Theta_1$  affine,  $\Theta_2$  affine  $\Rightarrow$  convexe

(3P) c) KKT:  $\Leftrightarrow \begin{cases} 2x_1 - 14 + \lambda_1 + \lambda_2 = 0 \\ 2x_2 - 6 + \lambda_1 + 2\lambda_2 = 0 \\ \lambda_1(x_1 + x_2 - 2) = \lambda_2(x_1 + 2x_2 - 3) = 0 \\ \lambda_1, \lambda_2 \geq 0 \\ x_1 + x_2 - 2 \leq 0, x_1 + 2x_2 - 3 \leq 0 \end{cases}$

1) Cas  $\lambda_1 = \lambda_2 = 0 \Rightarrow (\bar{x}_1, \bar{x}_2) = (7, 3) \notin \Omega$  Non

2) Cas  $x_1 + x_2 - 2 = 0, \lambda_2 = 0 \Rightarrow (\bar{x}_1, \bar{x}_2) = (3, -1)$ ,  $\lambda_1 = 8$  Oui

3) Cas  $\lambda_1 = 0, x_1 + 2x_2 - 3 = 0 \Rightarrow (\bar{x}_1, \bar{x}_2) = (5, -1) \notin \Omega$ ,  $\lambda_2 = 4$  Non

4)  $x_1 + x_2 - 2 = 0, x_1 + 2x_2 - 3 = 0 \Rightarrow (\bar{x}_1, \bar{x}_2) = (1, 1)$ ,  $\lambda_1 = 20, \lambda_2 = -9$  Non

donc la solution  $\bar{x} = (\bar{x}_1, \bar{x}_2) = (3, -1)$  et  $f(\bar{x}) = -33$

$$(1P) a) L(x, \lambda) = x_1^2 + x_2^2 - 14x_1 - 6x_2 - 7 + \lambda_1(x_1 + x_2 - 2) + \lambda_2(x_1 + 2x_2 - 3)$$

$$(2P) e). \sup_{\lambda \in \mathbb{R}_+^2} L(x, \lambda) = \begin{cases} +\infty & \text{si } x \notin \Omega \\ f(x) & \text{si } x \in \Omega \end{cases}$$

$$\inf_{x \in \mathbb{R}^2} \sup_{\lambda \in \mathbb{R}_+^2} L(x, \lambda) = \inf_{x \in \Omega} f(x) = f(\bar{x}) = f(3, -1) = -33 \quad (*).$$

$$(2P) b) \inf_{x \in \mathbb{R}^2} L(x, \lambda) = L(x^*, \lambda) \Rightarrow \begin{cases} \nabla_x L(x^*, \lambda) = 0 \Leftrightarrow \\ \begin{aligned} 2x_1^* - 14 + \lambda_1 + \lambda_2 &= 0 \\ 2x_2^* - 6 + \lambda_1 + 2\lambda_2 &= 0 \end{aligned} \end{cases}$$

$$\Leftrightarrow \begin{cases} x_1^* = -\frac{\lambda_1}{2} - \frac{\lambda_2}{2} + 7 \\ x_2^* = -\frac{\lambda_1}{2} + \lambda_2 + 3 \end{cases} \text{ et } L(x^*, \lambda) = -\frac{\lambda_1^2}{2} - \frac{5}{4}\lambda_2^2 - \frac{3}{2}\lambda_1\lambda_2 + 8\lambda_1 + \frac{\lambda_1 - \lambda_2}{2} - 65$$

$$\text{donc } \sup_{\lambda \in \mathbb{R}_+^2} \inf_{x \in \mathbb{R}^2} L(x, \lambda) = \sup_{\lambda \in \mathbb{R}_+^2} L(x^*, \lambda) = \sup_{\lambda \in \mathbb{R}_+^2} D(\lambda) \quad / D(\lambda) = L(x^*, \lambda) \text{ (le problème dual?)}$$

$$e) \sup_{\lambda \in \mathbb{R}_+^2} D(\lambda) = - \inf_{\lambda \in \mathbb{R}_+^2} -D(\lambda)$$

$$(2P) \text{ on résout le problème } \inf_{\lambda \geq 0} \left[ \frac{\lambda_1^2}{2} + \frac{5}{4}\lambda_2^2 + \frac{3}{2}\lambda_1\lambda_2 + 8\lambda_1 - \frac{\lambda_1 - \lambda_2}{2} + 65 \right]$$

$$\text{par KKT on trouve la solution } \bar{\lambda} = (\bar{\lambda}_1, \bar{\lambda}_2) = (8, 0) \text{ et } D(\bar{\lambda}) = -33.$$

$$\text{donc la gap de dual} = \text{val(P)} - \text{val(D)} = -33 + 33 = 0.$$