

## Solution Set 3

- 23. (a)**  $F(x, y) = -2x^2 + y^2 + 3xy - x - 2y$  is **jointly continuous** and twice continuously differentiable in  $(x, y)$ . Hence, it is strictly concave-convex in  $(x, y)$  iff  $\partial^2 F / \partial x^2 < 0$  and  $\partial^2 F / \partial y^2 > 0$ . Since,

$$\frac{\partial^2 F}{\partial x^2} = -4 < 0; \quad \frac{\partial^2 F}{\partial y^2} = 2 > 0,$$

**strict concavity-convexity** readily follows. Then, in view of the *saddle-point theorem* proved in class, the game admits a unique saddle point solution, since the interval  $[0, 1]$  is **compact**.

- (b)** The solution exists and is unique, as justified above.

Let  $\varphi$  and  $\psi$  be defined by [both map  $[0, 1]$  onto itself]

$$\begin{aligned} \max_x F(x, y) &= F(\varphi(y), y) & \forall y \in [0, 1] \\ \min_y F(x, y) &= F(x, \psi(x)) & \forall x \in [0, 1]. \end{aligned}$$

It readily follows that

$$\begin{aligned} \varphi(y) &= \begin{cases} (3y - 1)/4 & \text{if } y \geq \frac{1}{3} \\ 0 & \text{if } 0 \leq y < \frac{1}{3} \end{cases} \\ \psi(x) &= \begin{cases} (2 - 3x)/2 & \text{if } 0 \leq x < \frac{2}{3} \\ 0 & \text{if } x \geq \frac{2}{3} \end{cases}. \end{aligned}$$

Then, solving for  $x$  and  $y$  from  $\begin{cases} x = \varphi(y) \\ y = \psi(x) \end{cases}$ , we obtain the **unique** solution

$$\boxed{x^* = \frac{4}{17} \cong 0.235; \quad y^* = \frac{11}{17} \cong 0.647.}$$

The corresponding value of  $F$ , which is the *saddle-point value*, is  $F(x^*, y^*) \cong -0.7647$ .

- (c)** The sequence generated is

$$\begin{cases} x_{k+1} = \varphi(y_k) \\ y_{k+1} = \psi(x_{k+1}) \end{cases}$$

$\Rightarrow$

$$y_{k+1} = \psi(\varphi(y_k)) = \begin{cases} \frac{11}{8} - \frac{9}{8}y_k & y_k \geq \frac{1}{3} \\ 1 & y_k < \frac{1}{3} \end{cases}$$

Since the absolute value of the coefficient of  $y_k$  above is larger than 1, the sequence **does not converge**, regardless of what initial condition we choose for  $y_0$  (other than the equilibrium value:  $11/17$ ).

- 24. (a)** The proof here parallels that of the saddle-point equilibrium, given in class. As proven in class, the conditions given on  $X$ ,  $K_X$ ,  $Y$ ,  $K_Y$ , and  $F_1$  lead to the existence of an upper semi-continuous, closed, convex map  $g_1 : K_Y \rightarrow 2^{K_X}$  such that any  $x \in g_1(y)$  minimizes  $F_1(x, y)$  over  $x \in K_X$ , for each fixed  $y \in K_Y$ . Similarly, there exists an upper semi-continuous map  $g_2 : K_X \rightarrow 2^{K_Y}$ , which is also closed and convex, such that any  $y \in g_2(x)$  minimizes  $F_2(x, y)$  over  $y \in K_Y$ , for each fixed  $x \in K_X$ . Every Nash equilibrium is a common fixed point of these two point-to-set maps, and every such fixed point can be obtained from

$$x \in g_1(g_2(x)); \quad y \in g_2(x)$$

As shown in class, the composite map  $(g_1 \circ g_2)(\cdot) := g_1(g_2(\cdot))$  is a multifunction from  $K_X$  into  $2^{K_X}$ , which is also upper semi-continuous, closed and convex. Furthermore, since  $K_X$  is a compact subset of the metric space  $X$ ,  $(g_1 \circ g_2)$  has a fixed point, say  $x^*$ , by [Fan's Fixed-Point Theorem \(Correspondence #3, Theorem 9\)](#). Then, it readily follows that  $(x^*, y^* \in g_2(x^*))$  is a Nash equilibrium point.

**(b)** All conditions above hold here; in particular  $F_1$  is strictly convex in  $x$  and  $F_2$  is convex in  $y$ ; furthermore,  $K_X$  and  $K_Y$  are convex and compact. Hence, we know from the result above that there exists a Nash equilibrium. To compute it (and to determine whether it is unique or not), note that

$$g_1(y) = \frac{1}{y+1}, \quad g_2(x) = \begin{cases} 0 & \text{if } x > 1/2 \\ [0, 1] & \text{if } x = 1/2 \\ 1 & \text{if } x < 1/2 \end{cases} \Rightarrow (g_1 \circ g_2)(x) = \begin{cases} 1 & \text{if } x > 1/2 \\ [1/2, 1] & \text{if } x = 1/2 \\ 1/2 & \text{if } x < 1/2 \end{cases}$$

Hence,  $g_1$  is single-valued, but  $g_2$  as well as  $g_1 \circ g_2$  are multi-valued. It follows by inspection that  $g_1 \circ g_2$  has two fixed points:  $x = 1/2$  and  $x = 1$ . These correspond to  $y = 1$ , and  $y = 0$ , respectively. (Note that even though  $g_2(1/2) = [0, 1]$ , only the point  $y = 1$  in the interval  $[0, 1]$  leads to  $g_1(1) = 1/2$ .) Hence, the game admits two Nash equilibria:

$$x^* = \frac{1}{2}, y^* = 1 \quad \text{and} \quad x^\circ = 1, y^\circ = 0$$

- 25.** Let us check the four axioms:

- (i)  $\overline{(B, A)} = \overline{Tr[B^T Q \bar{A}]} = Tr[\bar{B}^T \bar{Q} A] = Tr[A^T \bar{Q}^T \bar{B}] = (A, B)$ , where the third equality holds because  $Tr[C] = Tr[C^T]$ , and the last one holds because  $Q$  is Hermitian.
- (ii)  $(A + B, C) = Tr[A^T Q \bar{C}] + Tr[B^T Q \bar{C}] = (A, C) + (B, C)$  – holds for all  $A$ ,  $B$  and  $C$ .
- (iii)  $(\lambda A, B) = \lambda Tr[A^T Q \bar{B}] = \lambda (A, B)$  – holds for all  $A$  and  $B$ .
- (iv)  $(A, A) = Tr[A^T Q \bar{A}] \rightarrow$  this is positive for all  $A \neq 0$  because the matrix  $Q$  is Hermitian and positive definite, i.e., all its eigenvalues are positive.

Hence,  $(A, B) = Tr[A^T Q \bar{B}]$  is indeed an inner product.

- 26. (a)** It is not an inner product on  $X$ , because

$$(x, x) = \left| \int_1^4 s^2 x(s) ds \right|^2$$

can be made zero without  $x$  being the zero function.

**(b) Yes**, it is an inner product on  $X$ , as all four axioms of an inner product are satisfied:

(i)  $(x, y) = (y, x)$  for all  $x, y \in X$ .

(ii)  $(x + y, z) = (x, z) + (y, z)$  for all  $x, y, z \in X$ .

(iii)  $(\lambda x, y) = \lambda(x, y)$  for all real numbers  $\lambda$  and all  $x, y \in X$ .

(iv)  $(x, x) = \int_1^4 t^3 x^2(t) dt \geq 0$  and is equal to zero iff  $x(t) = 0$  for all  $t \in [1, 4]$ , which is the zero element in  $X$ .

**27. (a)** Here the Hilbert space is  $H = L_2[-1, 2]$ , and  $M = \{m \in H : m(t) = a + bt, a, b \in \mathbf{R}\}$ . Note that  $M$  is a 2-dimensional subspace of  $H$ , which is also closed (it is in fact isomorphic to  $\mathbf{R}^2$ ). Hence the problem can be viewed as the optimization problem of minimizing  $\|x - m\|$  over  $m \in M$ , where  $x(t) = t^3$  is an element of  $H$ . The Projection Theorem directly applies here, leading to the conclusion that there exists a unique  $m_o \in M$  that solves this minimization problem, and that  $x - m_o \perp M$ .

**(b)** We have to solve for  $a$  and  $b$  from the two relationships:

$$x - m_o \perp 1 \Rightarrow (t^3 - m_o, 1) = 0 \quad \text{and} \quad x - m_o \perp t \Rightarrow (t^3 - m_o, t) = 0$$

Using  $m_o(t) = a + bt$ , we have  $\frac{5}{4} = a + \frac{1}{2}b$ ,  $\frac{21}{5} = \frac{1}{2}a + b \Rightarrow a = 0.2, b = 2.1$

**(c)** The minimum value of  $F$  is

$$\|x - m_o\|^2 = (x - m_o, x - m_o) = (x - m_o, x) = \|x\|^2 - (m_o, x) = \|x\|^2 - \|m_o\|^2$$

where  $\|x\|^2 = \frac{129}{7}$  and  $\|m_o\|^2 = 14.61 \Rightarrow \min_{m \in M} F(m) = F(m_o) = 3.8187$

**28.** Here the space  $L_2[-1, 2]$  is replaced by the same with only the inner product different:

$$(x, y)_2 = \int_{-1}^2 t^2 x(t) y(t) dt$$

This is also a Hilbert space (say,  $H$ ), and  $M$  as defined above is a closed subspace of  $H$ . Projection Theorem again applies, leading to existence of a unique solution. We again have to solve for  $a$  and  $b$  from the two relationships:

$$x - m_o \perp 1 \Rightarrow (t^3 - m_o, 1)_2 = 0 \quad \text{and} \quad x - m_o \perp t \Rightarrow (t^3 - m_o, t)_2 = 0$$

Substituting for  $m_o(t) = a + bt$ , and solving for  $a$  and  $b$ :  $a = 0.03361$ ,  $b = 2.773$ .

We have  $\|x\|^2 = \|t^3\|^2 = 56.778$  and  $\|m_o\|^2 = 51.547$

and hence the minimum value of  $F$  is:  $F(m_o) = \|x\|^2 - \|m_o\|^2 = 5.543$ .

**29.** Let  $y(t) = x(t)\sqrt{t}$ . Then, the problem is equivalent to one of minimizing  $\|y\|^2$  over  $H = L_2[1, 2]$ , subject to:  $(y, t^{-\frac{1}{6}}) = 1$  and  $(y, t^{\frac{1}{6}}) = -1$ , where the inner product  $(\cdot, \cdot)$  is the standard one on  $H$ .

Let  $Z$  be the closed subspace of  $H$  generated by  $\{t^{-\frac{1}{6}}, t^{\frac{1}{6}}\}$ . Then,  $H = Z \oplus Z^\perp$ , and hence any  $y \in H$  admits the unique additive decomposition  $y = z + m$ , where  $z \in Z$  and  $m \in M := Z^\perp$ . The constraints can then be written as  $(z, t^{-\frac{1}{6}}) = 1$  and  $(z, t^{\frac{1}{6}}) = -1$ , and the functional to be minimized is  $\|z\|^2 + \|m\|^2$  (since  $m \perp z$ ). But, since  $Z$  is two-dimensional (because  $t^{-\frac{1}{6}}$  and  $t^{\frac{1}{6}}$  are linearly independent), and we have two linearly independent constraints, the choice of an element out of  $Z$  that satisfies both constraints is unique. Letting  $y(t) = at^{-\frac{1}{6}} + bt^{\frac{1}{6}}$ , and solving for  $a$  and  $b$ , we have  $a = 486.3295$ ,  $b = -427.5050$ . Hence, the unique solution to the optimization problem is (in view of the relationship  $x(t) = y(t)/\sqrt{t}$ :

$$x^o(t) = 486.3295 t^{-\frac{2}{3}} - 427.5050 t^{-\frac{1}{3}}$$

- 30.** Choose the Hilbert space  $H = \mathbf{R}^n$ , with the inner product  $(x, y) = x^T Q y$ , where  $Q > 0$  (positive definite).

Let  $A^T = [a_1, a_2, \dots, a_m]$ ,  $b^T = [b_1, b_2, \dots, b_m]$ , where  $a_i$ 's are vectors and  $b_j$ 's are scalars. Further assume that  $a_1, \dots, a_m$  are linearly independent in  $\mathbf{R}^n$ , which is equivalent to saying that  $\text{rank}(A) = m$ .

Then, the constraint equation  $Ax = b$  can be written as

$$\begin{aligned} a_i^T x &= b_i, & i &= 1, \dots, m \\ \Leftrightarrow (x, \tilde{a}_i) &= b_i, & i &= 1, \dots, m \end{aligned}$$

where  $\tilde{a}_i = Q^{-1}a_i$ .

Note that since  $\det Q^{-1} \neq 0$ ,  $\{\tilde{a}_i\}_{i=1}^m$  is also a linearly independent set.

Hence, our optimization problem is

$$\text{minimize } \|x\| \quad \text{subject to } (x, \tilde{a}_i) = b_i, i = 1, \dots, m.$$

Applying Thm 2 (p. 65, Luenberger), which was discussed in class, we know that the solution is unique, and is given by

$$x_0 = \sum_{i=1}^m \lambda_i \tilde{a}_i = Q^{-1} A^T \lambda$$

where the  $\tilde{a}_i$ 's satisfy

$$\underbrace{\begin{bmatrix} (\tilde{a}_1, \tilde{a}_1) & (\tilde{a}_2, \tilde{a}_1) & \cdots & (\tilde{a}_m, \tilde{a}_1) \\ \vdots & & & \\ \vdots & & & \\ (\tilde{a}_1, \tilde{a}_m) & \cdots & \cdots & (\tilde{a}_m, \tilde{a}_m) \end{bmatrix}}_G \underbrace{\begin{bmatrix} \lambda_1 \\ \vdots \\ \vdots \\ \lambda_m \end{bmatrix}}_\lambda = \underbrace{\begin{bmatrix} b_1 \\ \vdots \\ \vdots \\ b_m \end{bmatrix}}_b.$$

The matrix  $G$  can easily be shown to be  $G = A Q^{-1} A^T$ , which is nonsingular because of the linear independence of columns of  $A^T$ . Then,  $\lambda$  is uniquely solved to yield

$$\lambda = G^{-1} b = (A Q^{-1} A^T)^{-1} b$$

and substituting this into the expression for  $x_0$ , we obtain:

$$\boxed{x_0 = Q^{-1}A^T(AQ^{-1}A^T)^{-1}b} \Rightarrow \text{unique solution.}$$

For those of you who have taken ECE 490, or are familiar with the contents of any nonlinear programming course, the parameter vector  $\lambda$  obtained above is the Lagrange multiplier, and the problem solved is the “quadratic programming problem with linear equality constraints”.

If  $A$  is not a full rank matrix, but the linear constraints are still consistent (compatible), then one reduces the number of equations by eliminating the redundant ones and arrive at a new matrix  $A$  which is now of full rank. Then, this new  $A$  would be used in the solution given above.

- 31.** Let  $Q$  be the space of polynomials on  $[a, b]$ , of degree  $n$  or less, and  $P$  be a subset of  $Q$ , consisting of elements,  $p$ , of  $Q$ , which further satisfy the constraint

$$(p, 1) = \int_a^b p(t) dt = 0,$$

that is, are orthogonal to the constant 1. Clearly, both  $Q$  and  $P$  are closed subspaces of  $L_2[a, b]$ , and  $P$  is further a subspace of  $Q$ . By the Projection Theorem, both minimization problems

$$\min_{p \in P} \|x - p\| \quad \text{and} \quad \min_{q \in Q} \|x - q\|$$

admit unique solutions, say  $p^o$  and  $q^o$ , respectively. Furthermore,  $x - p^o \perp P$  and  $x - q^o \perp Q$ , and  $p^o$  and  $q^o$  are the unique elements of  $P$  and  $Q$ , respectively, that satisfy these orthogonality relationships. By the same reasoning, the minimization problem  $\min_{p \in P} \|q^o - p\|$  also admits a unique solution; let us denote it by  $\hat{p}$ , and note that  $q^o - \hat{p} \perp P$ , with this relationship again satisfied uniquely by  $\hat{p}$ . Now, for any  $p \in P$ :

$$(x - \hat{p}, p) = (x - q^o + q^o - \hat{p}, p) = (x - q^o, p) + (q^o - \hat{p}, p) = 0$$

where the last result follows because  $q^o - \hat{p} \perp P$ , and  $x - q^o \perp Q \Rightarrow x - q^o \perp P \subset Q$ . But since  $p^o$  was the unique vector with the property  $x - p^o \perp P$ , it follows that  $\hat{p} = p^o$ .

- 32. i)** Let  $Y = \overline{[y_1, \dots, y_n]}$ . Write  $x \in X$  as  $x = x_y + x^\perp$  where  $x_y \in Y$ ,  $x^\perp \in Y^\perp$ . Then, because of orthogonality,  $\|x\|^2 = \|x_y + x^\perp\|^2 = \|x_y\|^2 + \|x^\perp\|^2$ . Let  $K = \{k \in Y : (k, y_i) \geq c_i\}$ . Then, since  $(x_y + x^\perp, y_i) = (x_y, y_i)$ ,

$$\inf_{\substack{x \in X \\ \exists (x, y_i) \geq c_i \\ i=1, \dots, n}} \|x\|^2 = \inf_{k \in K} \|k\|^2$$

$K$  is closed (because of  $\geq$ ) and convex (because  $(k, y_i)$  is a linear functional). Hence, what we have is a problem of *minimum distance to a convex set in a Hilbert space*. Thm 1 (p. 69 of Luenberger) applies with  $x = \theta$ , to ensure that  $\exists$  a unique  $k_0 \in K$  satisfying  $\inf_{k \in K} \|k\| = \|k_0\|$  and the solution is characterized by

$$(k_0, k_0 - k) \leq 0 \quad \forall k \in K \quad (1)$$

where  $k_0$  can be written as

$$k_0 = \sum_{i=1}^n a_i y_i \text{ for some } a_1, \dots, a_n. \quad (2)$$

*ii)* First note that, for  $k_0 = \sum_{j=1}^n a_j y_j \in K$ , we have  $(k_0, y_i) \geq c_i \Rightarrow \sum_{j=1}^n a_j (y_j, y_i) \geq c_i \Rightarrow$

$$\boxed{G^T a \geq c}$$

where  $G$  is the Gram matrix,  $c := (c_1, c_2, \dots, c_n)^T$  and  $a := (a_1, \dots, a_n)^T$ . Now we have to show that  $a \geq \theta$ . Toward this end, first note that for any fixed  $j$ , we can find a vector  $z_j$  such that

$$(z_j, y_j) > 0 \quad \text{and} \quad z_j \perp \overline{[y_1, y_2, \dots, y_{j-1}, y_{j+1}, \dots, y_n]}.$$

Choosing  $k = k_0 + z_j$ , clearly  $(k, y_i) = (k_0, y_i) + \underbrace{(z_j, y_i)}_{\geq 0} \geq \underbrace{(k_0, y_i)}_{\geq c_i} \Rightarrow k \in K$ .

Using this  $k$  in (1), we obtain  $-(k_0, z_j) \leq 0 \Rightarrow \sum_{i=1}^n a_i (y_i, z_j) \geq 0 \Rightarrow a_j (y_j, z_j) \geq 0$ . Since  $(y_j, z_j) > 0$  by construction, this says that  $a_j$  cannot be negative, and since  $j$  was arbitrary,

$$\boxed{a \geq \theta}$$

Now we prove the last part, i.e.,  $(k_0, y_i) > c_i \Rightarrow a_i = 0$ . Since  $\{y_j\}_{j=1}^n$  is a linearly independent set,

$$\exists z_i \text{ such that } z_i \perp \overline{[y_1, \dots, y_{i-1}, y_{i+1}, \dots, y_n]} \text{ and } (z_i, y_i) = -\epsilon < 0.$$

Let  $k = k_0 + z_i$ , which belongs to  $K$  if  $\epsilon$  is sufficiently small, since  $(k_0, y_i) > c_i$ .

Now, using this  $k$  in (1), we obtain

$$(k_0, -z_i) \leq 0 \Rightarrow \sum_{j=1}^n a_j (y_j, z_i) \geq 0 \Leftrightarrow a_i \underbrace{(y_i, z_i)}_{-\epsilon} \geq 0.$$

Since  $a_i \geq 0$ , this is possible only if  $a_i = 0$ . Hence,

$$\boxed{a_i = 0 \text{ if } (k_0, y_i) > c_i}$$

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