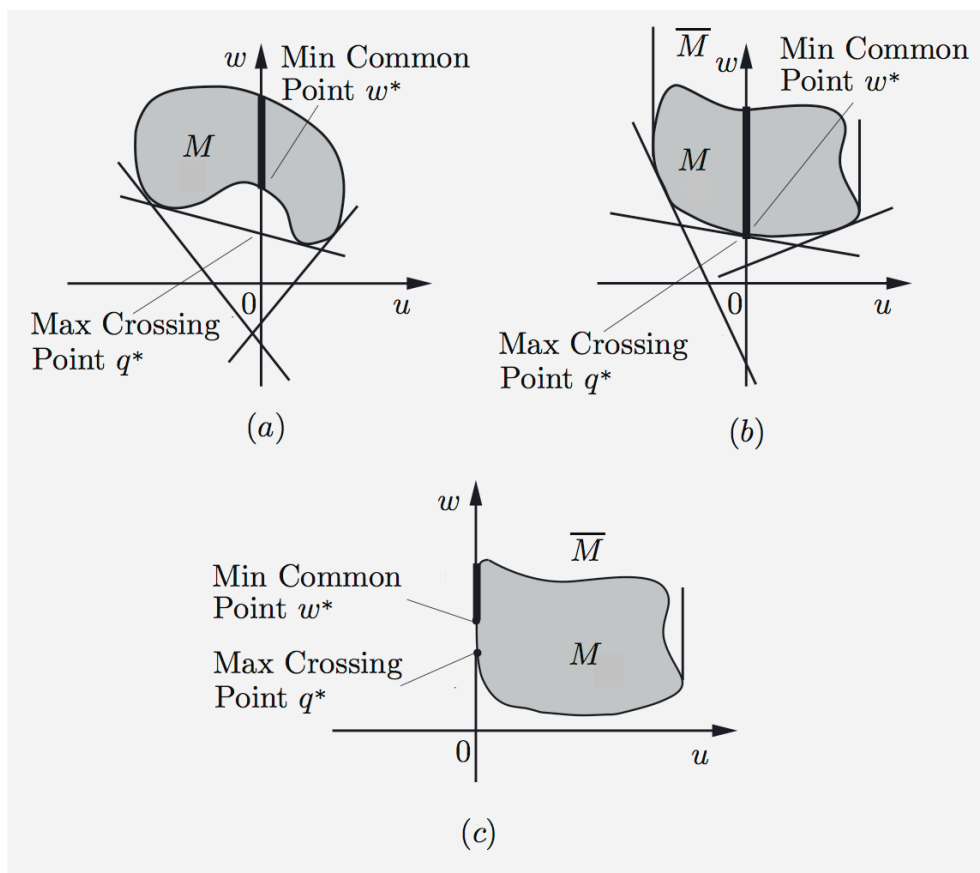


LECTURE 9

LECTURE OUTLINE

- Minimax problems and zero-sum games
- Min Common / Max Crossing duality for minimax and zero-sum games
- Min Common / Max Crossing duality theorems
- Strong duality conditions
- Existence of dual optimal solutions

Reading: Sections 3.4, 4.3, 4.4, 5.1



REVIEW OF THE MC/MC FRAMEWORK

- Given set $M \subset \mathfrak{R}^{n+1}$,

$$w^* = \inf_{(0,w) \in M} w, \quad q^* = \sup_{\mu \in \mathfrak{R}^n} q(\mu) \triangleq \inf_{(u,w) \in M} \{w + \mu' u\}$$

- **Weak Duality:** $q^* \leq w^*$
- **Important special case:** $M = \text{epi}(p)$. Then $w^* = p(0)$, $q^* = p^{**}(0)$, so we have $w^* = q^*$ if p is closed, proper, convex.
- Some applications:
 - Constrained optimization: $\min_{x \in X, g(x) \leq 0} f(x)$,
with $p(u) = \inf_{x \in X, g(x) \leq u} f(x)$
 - Other optimization problems: Fenchel and conic optimization
 - Useful theorems related to optimization: Farkas' lemma, theorems of the alternative
 - Subgradient theory
 - Minimax problems, 0-sum games
- **Strong Duality:** $q^* = w^*$. Requires that M have some convexity structure, among other conditions

MINIMAX PROBLEMS

Given $\phi : X \times Z \mapsto \mathfrak{R}$, where $X \subset \mathfrak{R}^n$, $Z \subset \mathfrak{R}^m$
consider

$$\begin{aligned} & \text{minimize} && \sup_{z \in Z} \phi(x, z) \\ & \text{subject to} && x \in X \end{aligned}$$

or

$$\begin{aligned} & \text{maximize} && \inf_{x \in X} \phi(x, z) \\ & \text{subject to} && z \in Z. \end{aligned}$$

- Some important contexts:
 - Constrained optimization duality theory
 - Zero sum game theory
- We always have

$$\sup_{z \in Z} \inf_{x \in X} \phi(x, z) \leq \inf_{x \in X} \sup_{z \in Z} \phi(x, z)$$

- **Key question:** When does equality hold?

CONSTRAINED OPTIMIZATION DUALITY

- For the problem

$$\begin{aligned} & \text{minimize } f(x) \\ & \text{subject to } x \in X, \quad g(x) \leq 0 \end{aligned}$$

introduce the Lagrangian function

$$L(x, \mu) = f(x) + \mu'g(x)$$

- Primal problem (equivalent to the original)

$$\min_{x \in X} \sup_{\mu \geq 0} L(x, \mu) = \begin{cases} f(x) & \text{if } g(x) \leq 0, \\ \infty & \text{otherwise,} \end{cases}$$

- Dual problem

$$\max_{\mu \geq 0} \inf_{x \in X} L(x, \mu)$$

- Key duality question: Is it true that

$$\inf_{x \in \mathfrak{R}^n} \sup_{\mu \geq 0} L(x, \mu) = w^* \stackrel{?}{=} q^* = \sup_{\mu \geq 0} \inf_{x \in \mathfrak{R}^n} L(x, \mu)$$

ZERO SUM GAMES

- Two players: 1st chooses $i \in \{1, \dots, n\}$, 2nd chooses $j \in \{1, \dots, m\}$.
- If i and j are selected, the 1st player gives a_{ij} to the 2nd.
- Mixed strategies are allowed: The two players select probability distributions

$$x = (x_1, \dots, x_n), \quad z = (z_1, \dots, z_m)$$

over their possible choices.

- Probability of (i, j) is $x_i z_j$, so the expected amount to be paid by the 1st player

$$x'Az = \sum_{i,j} a_{ij} x_i z_j$$

where A is the $n \times m$ matrix with elements a_{ij} .

- Each player optimizes his choice against the worst possible selection by the other player. So
 - 1st player minimizes $\max_z x'Az$
 - 2nd player maximizes $\min_x x'Az$

SADDLE POINTS

Definition: (x^*, z^*) is called a *saddle point* of ϕ if

$$\phi(x^*, z) \leq \phi(x^*, z^*) \leq \phi(x, z^*), \quad \forall x \in X, \forall z \in Z$$

Proposition: (x^*, z^*) is a saddle point if and only if the minimax equality holds and

$$x^* \in \arg \min_{x \in X} \sup_{z \in Z} \phi(x, z), \quad z^* \in \arg \max_{z \in Z} \inf_{x \in X} \phi(x, z) \quad (*)$$

Proof: If (x^*, z^*) is a saddle point, then

$$\begin{aligned} \inf_{x \in X} \sup_{z \in Z} \phi(x, z) &\leq \sup_{z \in Z} \phi(x^*, z) = \phi(x^*, z^*) \\ &= \inf_{x \in X} \phi(x, z^*) \leq \sup_{z \in Z} \inf_{x \in X} \phi(x, z) \end{aligned}$$

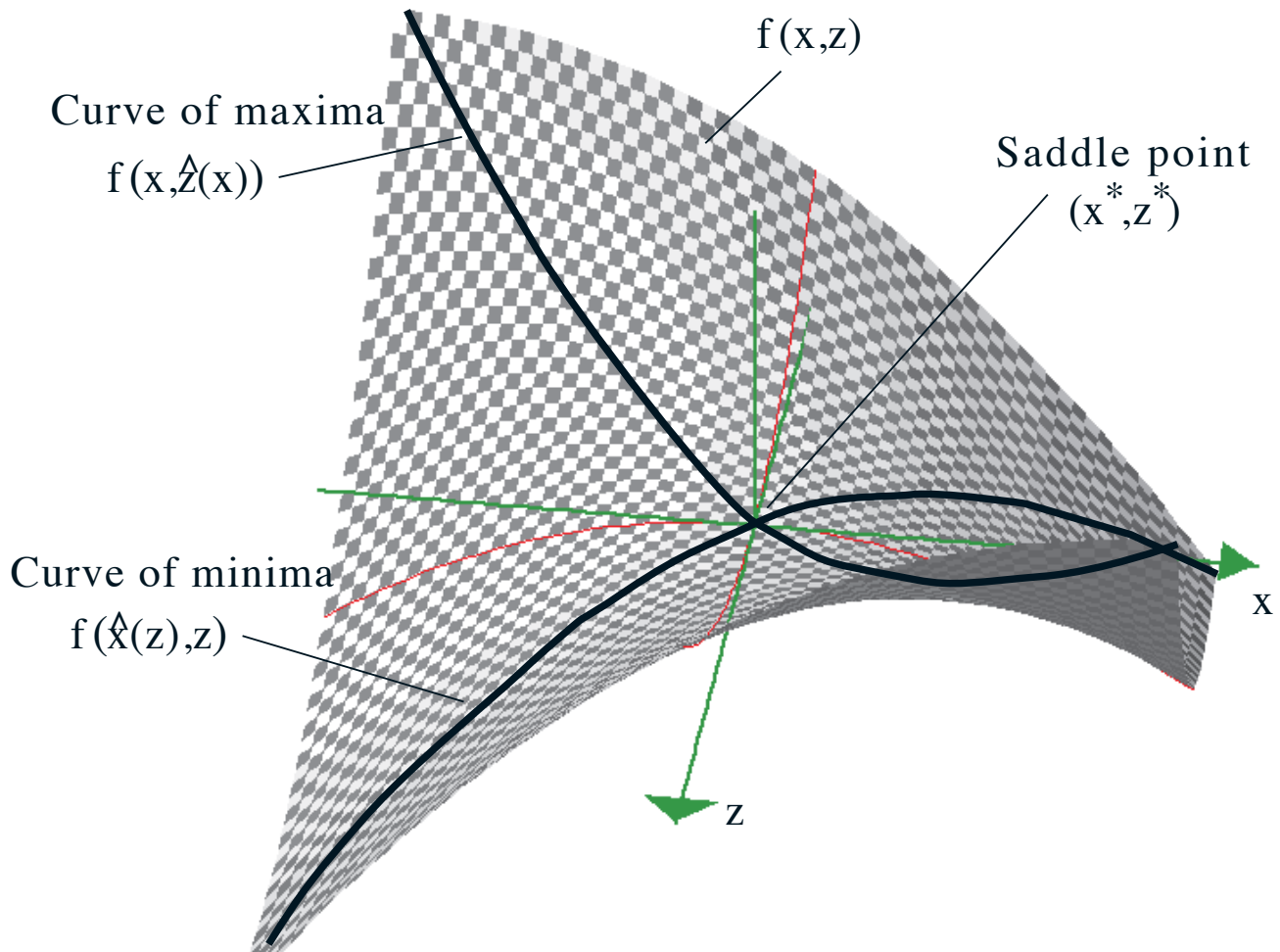
By the minimax inequality, the above holds as an equality throughout, so the minimax equality and Eq. (*) hold.

Conversely, if Eq. (*) holds, then

$$\begin{aligned} \sup_{z \in Z} \inf_{x \in X} \phi(x, z) &= \inf_{x \in X} \phi(x, z^*) \leq \phi(x^*, z^*) \\ &\leq \sup_{z \in Z} \phi(x^*, z) = \inf_{x \in X} \sup_{z \in Z} \phi(x, z) \end{aligned}$$

Using the minimax equ., (x^*, z^*) is a saddle point.

VISUALIZATION



The curve of maxima $f(x, \hat{z}(x))$ lies above the curve of minima $f(\hat{x}(z), z)$, where

$$\hat{z}(x) = \arg \max_z f(x, z), \quad \hat{x}(z) = \arg \min_x f(x, z)$$

Saddle points correspond to points where these two curves meet.

MINIMAX MC/MC FRAMEWORK

- Introduce perturbation function $p : \mathfrak{R}^m \mapsto [-\infty, \infty]$

$$p(u) = \inf_{x \in X} \sup_{z \in Z} \{ \phi(x, z) - u'z \}, \quad u \in \mathfrak{R}^m$$

- Apply the MC/MC framework with $M = \text{epi}(p)$. If p is convex, closed, and proper, no duality gap.
- Introduce $\hat{\text{cl}} \phi$, the *concave closure* of ϕ viewed as a function of z for fixed x
- We have

$$\sup_{z \in Z} \phi(x, z) = \sup_{z \in \mathfrak{R}^m} (\hat{\text{cl}} \phi)(x, z),$$

so

$$w^* = p(0) = \inf_{x \in X} \sup_{z \in \mathfrak{R}^m} (\hat{\text{cl}} \phi)(x, z).$$

- The dual function can be shown to be

$$q(\mu) = \inf_{x \in X} (\hat{\text{cl}} \phi)(x, \mu), \quad \forall \mu \in \mathfrak{R}^m$$

so if $\phi(x, \cdot)$ is concave and closed,

$$w^* = \inf_{x \in X} \sup_{z \in \mathfrak{R}^m} \phi(x, z), \quad q^* = \sup_{z \in \mathfrak{R}^m} \inf_{x \in X} \phi(x, z)$$

PROOF OF FORM OF DUAL FUNCTION

- Write $p(u) = \inf_{x \in X} p_x(u)$, where

$$p_x(u) = \sup_{z \in Z} \{ \phi(x, z) - u'z \}, \quad x \in X,$$

and note that

$$\inf_{u \in \mathfrak{R}^m} \{ p_x(u) + u' \mu \} = - \sup_{u \in \mathfrak{R}^m} \{ u'(-\mu) - p_x(u) \} = -p_x^*(-\mu)$$

Except for a sign change, p_x is the conjugate of $(-\phi)(x, \cdot)$ [assuming $(-\hat{\text{cl}} \phi)(x, \cdot)$ is proper], so

$$p_x^*(-\mu) = -(\hat{\text{cl}} \phi)(x, \mu).$$

Hence, for all $\mu \in \mathfrak{R}^m$,

$$\begin{aligned} q(\mu) &= \inf_{u \in \mathfrak{R}^m} \{ p(u) + u' \mu \} \\ &= \inf_{u \in \mathfrak{R}^m} \inf_{x \in X} \{ p_x(u) + u' \mu \} \\ &= \inf_{x \in X} \inf_{u \in \mathfrak{R}^m} \{ p_x(u) + u' \mu \} \\ &= \inf_{x \in X} \{ -p_x^*(-\mu) \} \\ &= \inf_{x \in X} (\hat{\text{cl}} \phi)(x, \mu) \end{aligned}$$

DUALITY THEOREMS

- Assume that $w^* < \infty$ and that the set

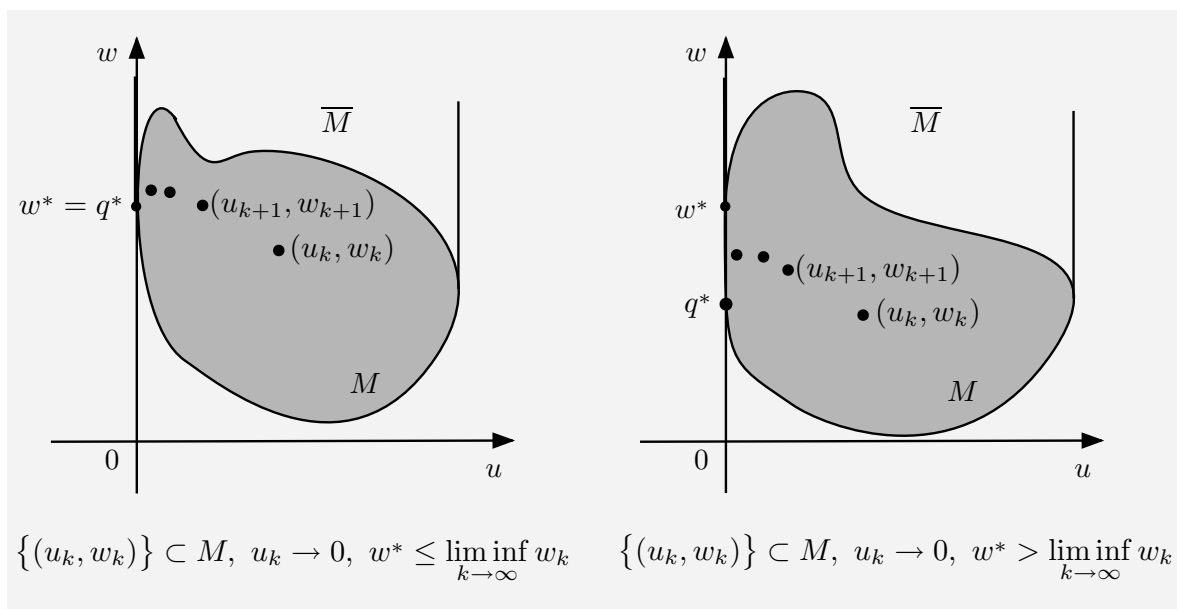
$$\bar{M} = \left\{ (u, w) \mid \text{there exists } \bar{w} \text{ with } \bar{w} \leq w \text{ and } (u, \bar{w}) \in M \right\}$$

is convex.

- **Min Common/Max Crossing Theorem I:**

We have $q^* = w^*$ if and only if for every sequence $\{(u_k, w_k)\} \subset M$ with $u_k \rightarrow 0$, there holds

$$w^* \leq \liminf_{k \rightarrow \infty} w_k.$$



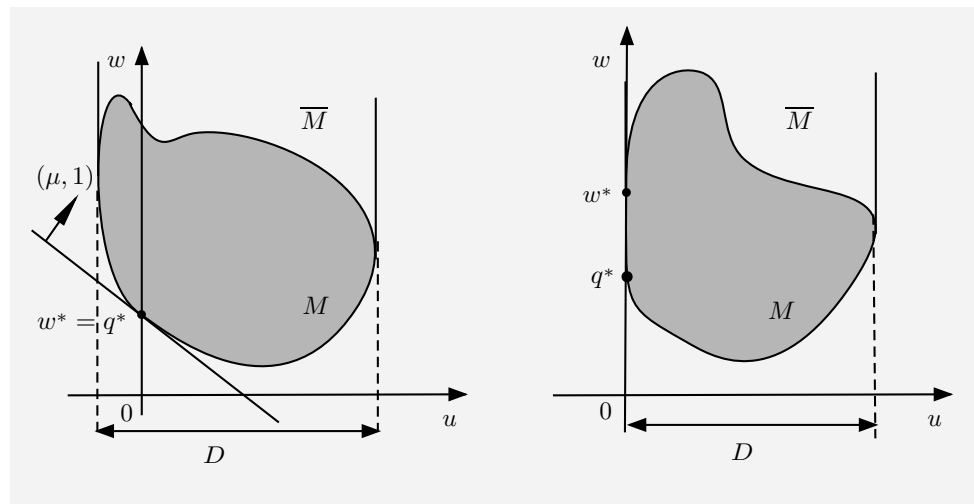
- **Corollary:** If $M = \text{epi}(p)$ where p is closed proper convex and $p(0) < \infty$, then $q^* = w^*$.

DUALITY THEOREMS (CONTINUED)

- **Min Common/Max Crossing Theorem II:** Assume in addition that $-\infty < w^*$ and that

$$D = \{u \mid \text{there exists } w \in \mathfrak{R} \text{ with } (u, w) \in \overline{M}\}$$

contains the origin in its relative interior. Then $q^* = w^*$ and there exists μ such that $q(\mu) = q^*$.



- Furthermore, the set $\{\mu \mid q(\mu) = q^*\}$ is nonempty and compact if and only if D contains the origin in its interior.
- **Min Common/Max Crossing Theorem III:** Involves polyhedral assumptions, and will be developed later.

PROOF OF THEOREM I

- Assume that $q^* = w^*$. Let $\{(u_k, w_k)\} \subset M$ be such that $u_k \rightarrow 0$. Then,

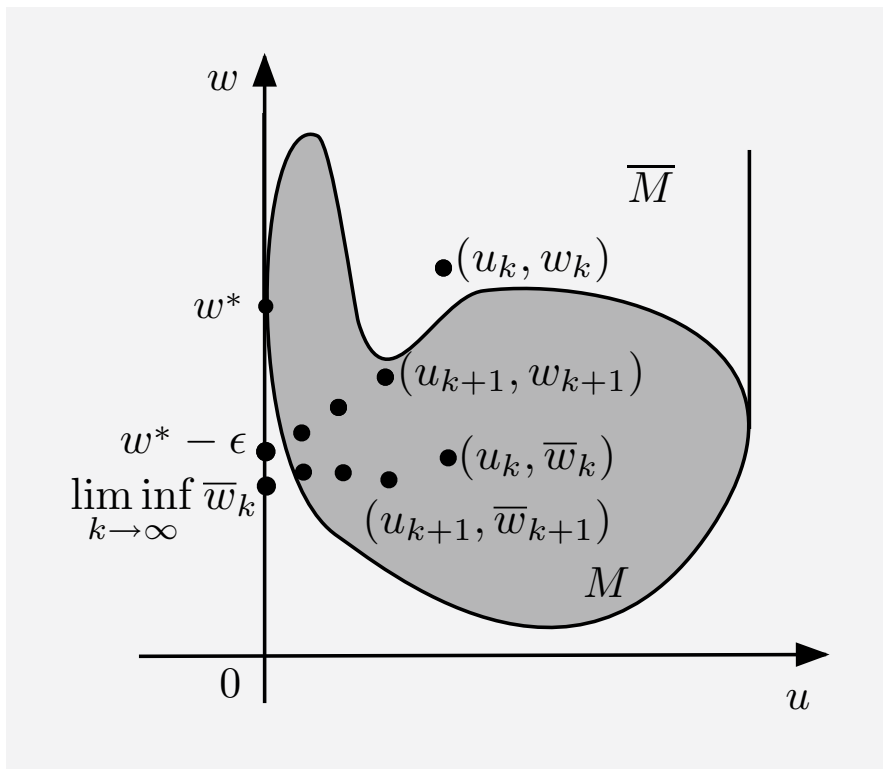
$$q(\mu) = \inf_{(u,w) \in M} \{w + \mu' u\} \leq w_k + \mu' u_k, \quad \forall k, \forall \mu \in \mathfrak{R}^n$$

Taking the limit as $k \rightarrow \infty$, we obtain $q(\mu) \leq \liminf_{k \rightarrow \infty} w_k$, for all $\mu \in \mathfrak{R}^n$, implying that

$$w^* = q^* = \sup_{\mu \in \mathfrak{R}^n} q(\mu) \leq \liminf_{k \rightarrow \infty} w_k$$

Conversely, assume that for every sequence $\{(u_k, w_k)\} \subset M$ with $u_k \rightarrow 0$, there holds $w^* \leq \liminf_{k \rightarrow \infty} w_k$. If $w^* = -\infty$, then $q^* = -\infty$, by weak duality, so assume that $-\infty < w^*$. Steps:

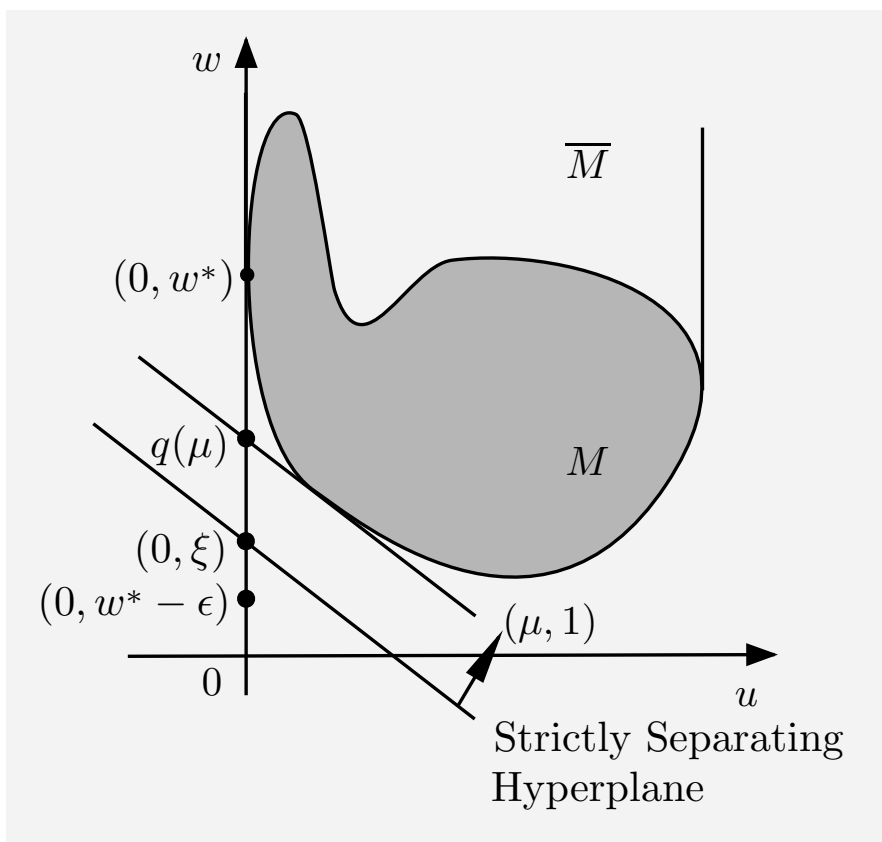
- **Step 1:** $(0, w^* - \epsilon) \notin \text{cl}(\overline{M})$ for any $\epsilon > 0$.



PROOF OF THEOREM I (CONTINUED)

• **Step 2:** \overline{M} does not contain any vertical lines. If this were not so, $(0, -1)$ would be a direction of recession of $\text{cl}(\overline{M})$. Because $(0, w^*) \in \text{cl}(\overline{M})$, the entire halfline $\{(0, w^* - \epsilon) \mid \epsilon \geq 0\}$ belongs to $\text{cl}(\overline{M})$, contradicting Step 1.

• **Step 3:** For any $\epsilon > 0$, since $(0, w^* - \epsilon) \notin \text{cl}(\overline{M})$, there exists a nonvertical hyperplane strictly separating $(0, w^* - \epsilon)$ and \overline{M} . This hyperplane crosses the $(n + 1)$ st axis at a vector $(0, \xi)$ with $w^* - \epsilon \leq \xi \leq w^*$, so $w^* - \epsilon \leq q^* \leq w^*$. Since ϵ can be arbitrarily small, it follows that $q^* = w^*$.



PROOF OF THEOREM II

- Note that $(0, w^*)$ is not a relative interior point of \overline{M} . Therefore, by the Proper Separation Theorem, there is a hyperplane that passes through $(0, w^*)$, contains \overline{M} in one of its closed halfspaces, but does not fully contain \overline{M} , i.e., for some $(\mu, \beta) \neq (0, 0)$

$$\beta w^* \leq \mu' u + \beta w, \quad \forall (u, w) \in \overline{M},$$

$$\beta w^* < \sup_{(u, w) \in \overline{M}} \{\mu' u + \beta w\}$$

Will show that the hyperplane is nonvertical.

- Since for any $(\bar{u}, \bar{w}) \in M$, the set \overline{M} contains the halfline $\{(\bar{u}, w) \mid \bar{w} \leq w\}$, it follows that $\beta \geq 0$. If $\beta = 0$, then $0 \leq \mu' u$ for all $u \in D$. Since $0 \in \text{ri}(D)$ by assumption, we must have $\mu' u = 0$ for all $u \in D$ a contradiction. Therefore, $\beta > 0$, and we can assume that $\beta = 1$. It follows that

$$w^* \leq \inf_{(u, w) \in \overline{M}} \{\mu' u + w\} = q(\mu) \leq q^*$$

Since the inequality $q^* \leq w^*$ holds always, we must have $q(\mu) = q^* = w^*$.

NONLINEAR FARKAS' LEMMA

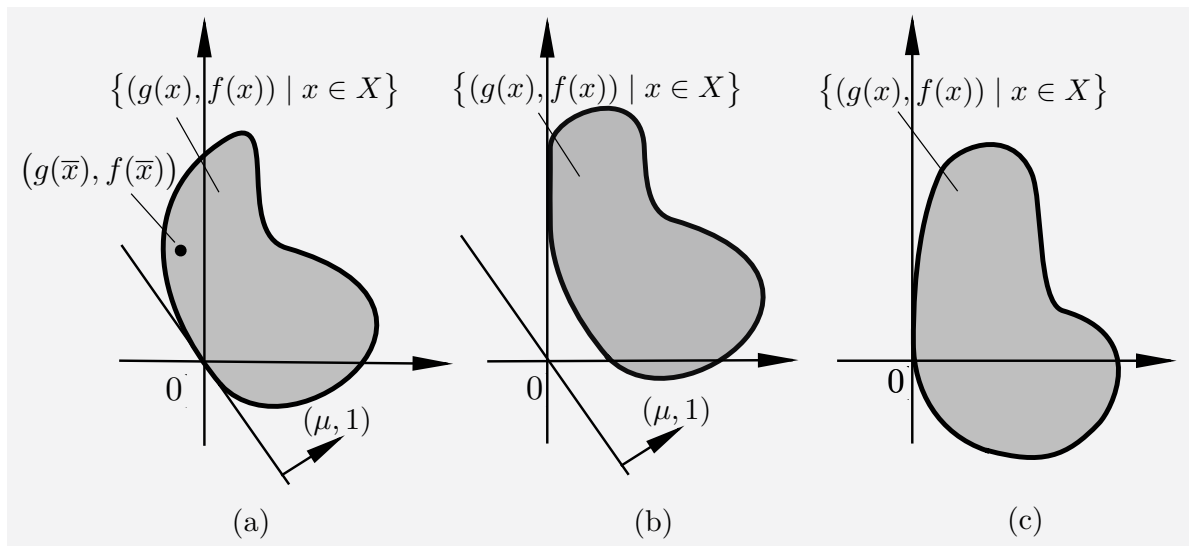
- Let $X \subset \mathfrak{R}^n$, $f : X \mapsto \mathfrak{R}$, and $g_j : X \mapsto \mathfrak{R}$, $j = 1, \dots, r$, be convex. Assume that

$$f(x) \geq 0, \quad \forall x \in X \text{ with } g(x) \leq 0$$

Let

$$Q^* = \{ \mu \mid \mu \geq 0, f(x) + \mu'g(x) \geq 0, \forall x \in X \}.$$

Then Q^* is nonempty and compact if and only if there exists a vector $\bar{x} \in X$ such that $g_j(\bar{x}) < 0$ for all $j = 1, \dots, r$.



- The lemma asserts the existence of a nonvertical hyperplane in \mathfrak{R}^{r+1} , with normal $(\mu, 1)$, that passes through the origin and contains the set

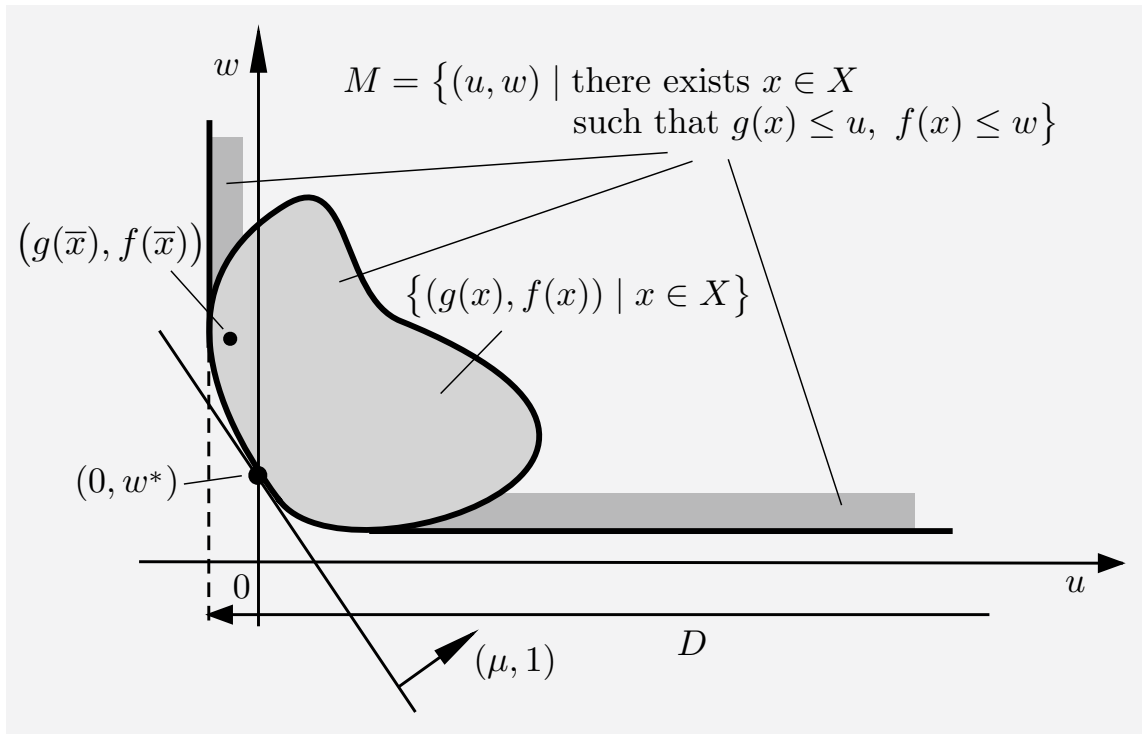
$$\{ (g(x), f(x)) \mid x \in X \}$$

in its positive halfspace.

PROOF OF NONLINEAR FARKAS' LEMMA

- Apply MC/MC to

$$M = \{(u, w) \mid \text{there is } x \in X \text{ s. t. } g(x) \leq u, f(x) \leq w\}$$



- M is equal to \overline{M} and is formed as the union of positive orthants translated to points $(g(x), f(x))$, $x \in X$.
- The convexity of X , f , and g_j implies convexity of M .
- MC/MC Theorem II applies: we have

$$D = \{u \mid \text{there exists } w \in \Re \text{ with } (u, w) \in \overline{M}\}$$

and $0 \in \text{int}(D)$, because $(g(\bar{x}), f(\bar{x})) \in M$.