

6.252 NONLINEAR PROGRAMMING

LECTURE 13: INEQUALITY CONSTRAINTS

LECTURE OUTLINE

- Inequality Constrained Problems
- Necessary Conditions
- Sufficiency Conditions
- Linear Constraints

Inequality constrained problem

minimize $f(x)$

subject to $h(x) = 0, \quad g(x) \leq 0$

where $f : \mathbb{R}^n \mapsto \mathbb{R}$, $h : \mathbb{R}^n \mapsto \mathbb{R}^m$, $g : \mathbb{R}^n \mapsto \mathbb{R}^r$ are continuously differentiable. Here

$$h = (h_1, \dots, h_m), \quad g = (g_1, \dots, g_r).$$

TREATING INEQUALITIES AS EQUATIONS

- Consider the *set of active inequality constraints*

$$A(x) = \{j \mid g_j(x) = 0\}.$$

- If x^* is a local minimum:
 - The active inequality constraints at x^* can be treated as equations
 - The inactive constraints at x^* don't matter
- Assuming regularity of x^* and assigning zero Lagrange multipliers to inactive constraints,

$$\nabla f(x^*) + \sum_{i=1}^m \lambda_i^* \nabla h_i(x^*) + \sum_{j=1}^r \mu_j^* \nabla g_j(x^*) = 0,$$

$$\mu_j^* = 0, \quad \forall j \notin A(x^*).$$

- Extra property: $\mu_j^* \geq 0$ for all j .

Intuitive reason: Relax j th constraint, $g_j(x) \leq u_j$,

$$\mu_j^* = -(\Delta \text{cost due to } u_j) / u_j$$

BASIC RESULTS

Kuhn-Tucker Necessary Conditions: Let x^* be a local minimum and a regular point. Then there exist unique Lagrange mult. vectors $\lambda^* = (\lambda_1^*, \dots, \lambda_m^*)$, $\mu^* = (\mu_1^*, \dots, \mu_r^*)$, such that

$$\nabla_x L(x^*, \lambda^*, \mu^*) = 0,$$

$$\mu_j^* \geq 0, \quad j = 1, \dots, r,$$

$$\mu_j^* = 0, \quad \forall j \notin A(x^*).$$

If f , h , and g are twice cont. differentiable,

$$y' \nabla_{xx}^2 L(x^*, \lambda^*, \mu^*) y \geq 0, \quad \text{for all } y \in V(x^*),$$

where

$$V(x^*) = \{y \mid \nabla h(x^*)' y = 0, \nabla g_j(x^*)' y = 0, j \in A(x^*)\}.$$

• Similar sufficiency conditions and sensitivity results. They require strict complementarity, i.e.,

$$\mu_j^* > 0, \quad \forall j \in A(x^*).$$

PROOF OF KUHN-TUCKER CONDITIONS

Use equality-constraints result to obtain all the conditions except for $\mu_j^* \geq 0$ for $j \in A(x^*)$. Introduce the penalty functions

$$g_j^+(x) = \max\{0, g_j(x)\}, \quad j = 1, \dots, r,$$

and for $k = 1, 2, \dots$, let x^k minimize

$$f(x) + \frac{k}{2} \|h(x)\|^2 + \frac{k}{2} \sum_{j=1}^r (g_j^+(x))^2 + \frac{1}{2} \|x - x^*\|^2$$

over a closed sphere of x such that $f(x^*) \leq f(x)$. Using the same argument as for equality constraints,

$$\lambda_i^* = \lim_{k \rightarrow \infty} kh_i(x^k), \quad i = 1, \dots, m,$$

$$\mu_j^* = \lim_{k \rightarrow \infty} kg_j^+(x^k), \quad j = 1, \dots, r.$$

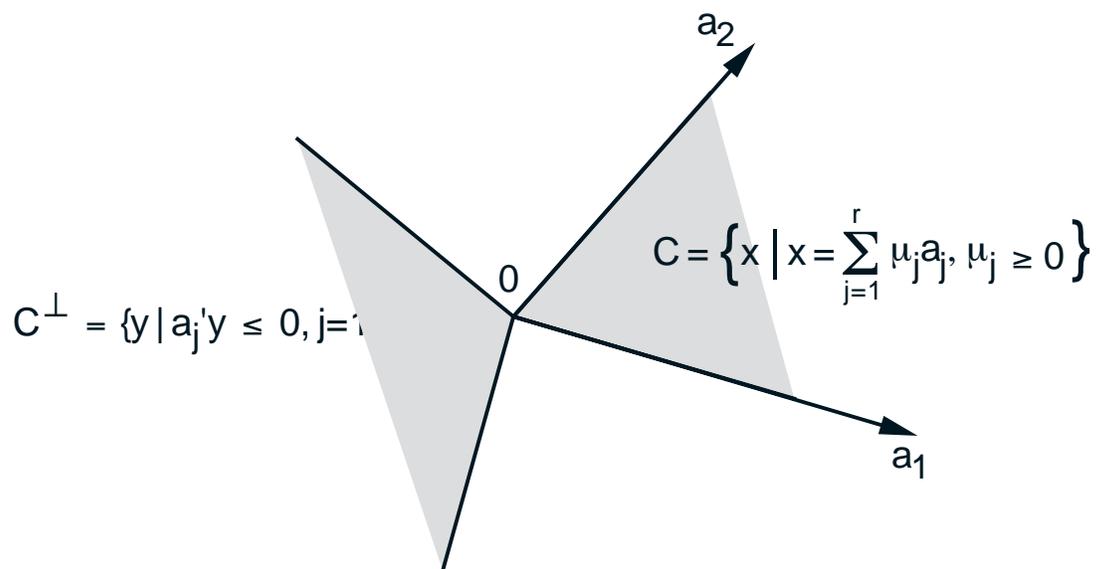
Since $g_j^+(x^k) \geq 0$, we obtain $\mu_j^* \geq 0$ for all j .

LINEAR CONSTRAINTS

- Consider the problem $\min_{a'_j x \leq b_j, j=1, \dots, r} f(x)$.
- Remarkable property: No need for regularity.
- Proposition: If x^* is a local minimum, there exist μ_1^*, \dots, μ_r^* with $\mu_j^* \geq 0, j = 1, \dots, r$, such that

$$\nabla f(x^*) + \sum_{j=1}^r \mu_j^* a_j = 0, \quad \mu_j^* = 0, \quad \forall j \notin A(x^*).$$

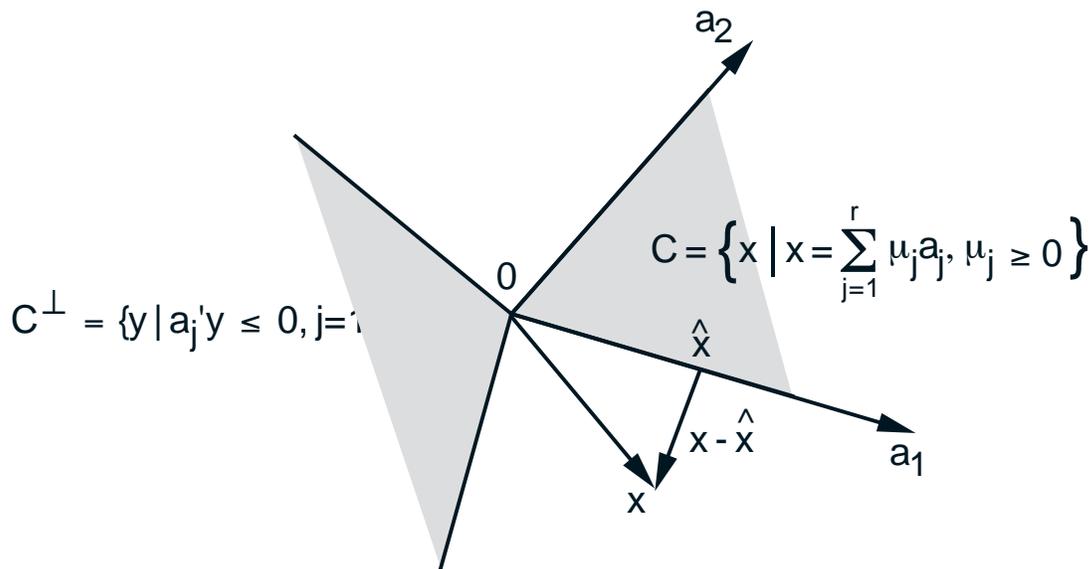
- Proof uses Farkas Lemma: Consider the cones C and C^\perp



$$x \in C \quad \text{iff} \quad x' y \leq 0, \quad \forall y \in C^\perp.$$

PROOF OF FARKAS LEMMA

$$x \in C \quad \text{iff} \quad x'y \leq 0, \quad \forall y \in C^\perp.$$



Proof: First show that C is closed (nontrivial). Then, let x be such that $x'y \leq 0, \forall y \in C^\perp$, and consider its projection \hat{x} on C . We have

$$x'(x - \hat{x}) = \|x - \hat{x}\|^2, \quad (*)$$

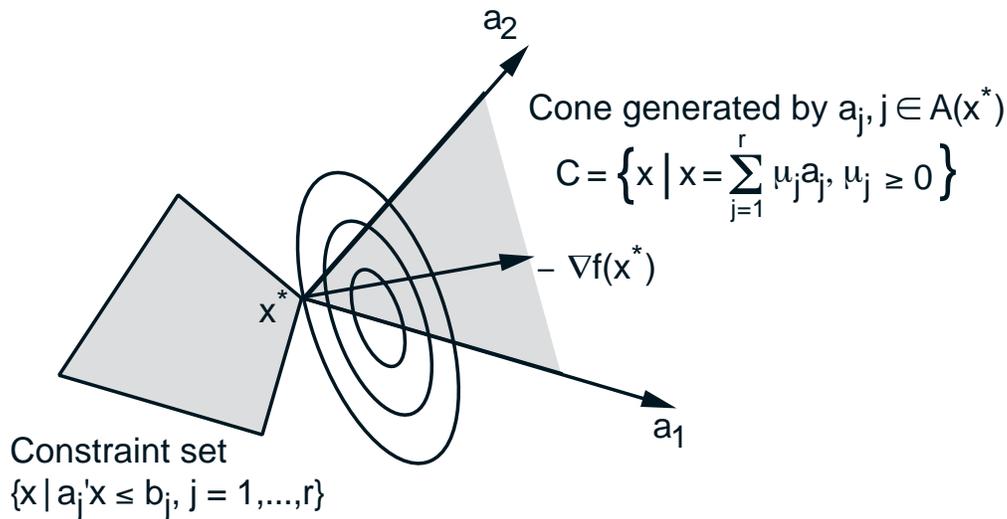
$$(x - \hat{x})'a_j \leq 0, \quad \forall j.$$

Hence, $(x - \hat{x}) \in C^\perp$, and using the hypothesis,

$$x'(x - \hat{x}) \leq 0. \quad (**)$$

From $(*)$ and $(**)$, we obtain $x = \hat{x}$, so $x \in C$.

PROOF OF LAGRANGE MULTIPLIER RESULT



The local min x^* of the original problem is also a local min for the problem $\min_{a_j'x \leq b_j, j \in A(x^*)} f(x)$. Hence

$$\nabla f(x^*)'(x - x^*) \geq 0, \quad \forall x \text{ with } a_j'x \leq b_j, j \in A(x^*).$$

Since a constraint $a_j'x \leq b_j, j \in A(x^*)$ can also be expressed as $a_j'(x - x^*) \leq 0$, we have

$$\nabla f(x^*)'y \geq 0, \quad \forall y \text{ with } a_j'y \leq 0, j \in A(x^*).$$

From Farkas' lemma, $-\nabla f(x^*)$ has the form

$$\sum_{j \in A(x^*)} \mu_j^* a_j, \quad \text{for some } \mu_j^* \geq 0, j \in A(x^*).$$

Let $\mu_j^* = 0$ for $j \notin A(x^*)$.

CONVEX COST AND LINEAR CONSTRAINTS

Let $f : \mathbb{R}^n \mapsto \mathbb{R}$ be convex and cont. differentiable, and let J be a subset of the index set $\{1, \dots, r\}$. Then x^* is a global minimum for the problem

$$\text{minimize } f(x)$$

$$\text{subject to } a'_j x \leq b_j, \quad j = 1, \dots, r,$$

if and only if x^* is feasible and there exist scalars μ_j^* , $j \in J$, such that

$$\mu_j^* \geq 0, \quad j \in J,$$

$$\mu_j^* = 0, \quad \forall j \in J \text{ with } j \notin A(x^*),$$

$$x^* = \arg \min_{\substack{a'_j x \leq b_j \\ j \notin J}} \left\{ f(x) + \sum_{j \in J} \mu_j^* (a'_j x - b_j) \right\}.$$

- Proof is immediate if $J = \{1, \dots, r\}$.
- Example: Simplex Constraint.