

LECTURE 21NONLINEAR OPTIMIZATION WITH
INEQUALITY CONSTRAINTS (PART II)

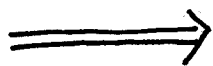
.. (NIP)

$$\begin{aligned} & \text{minimize } f(x) \\ & x \in \mathbb{R}^n \\ & \text{subject to} \\ & c_j(x) \geq 0 \quad j=1, \dots, m \end{aligned}$$

FIRST ORDER OPTIMALITY CONDITION FOR NIP

Suppose x_* is a point where constraint qualification holds.

x_* is a local minimizer of (NIP)



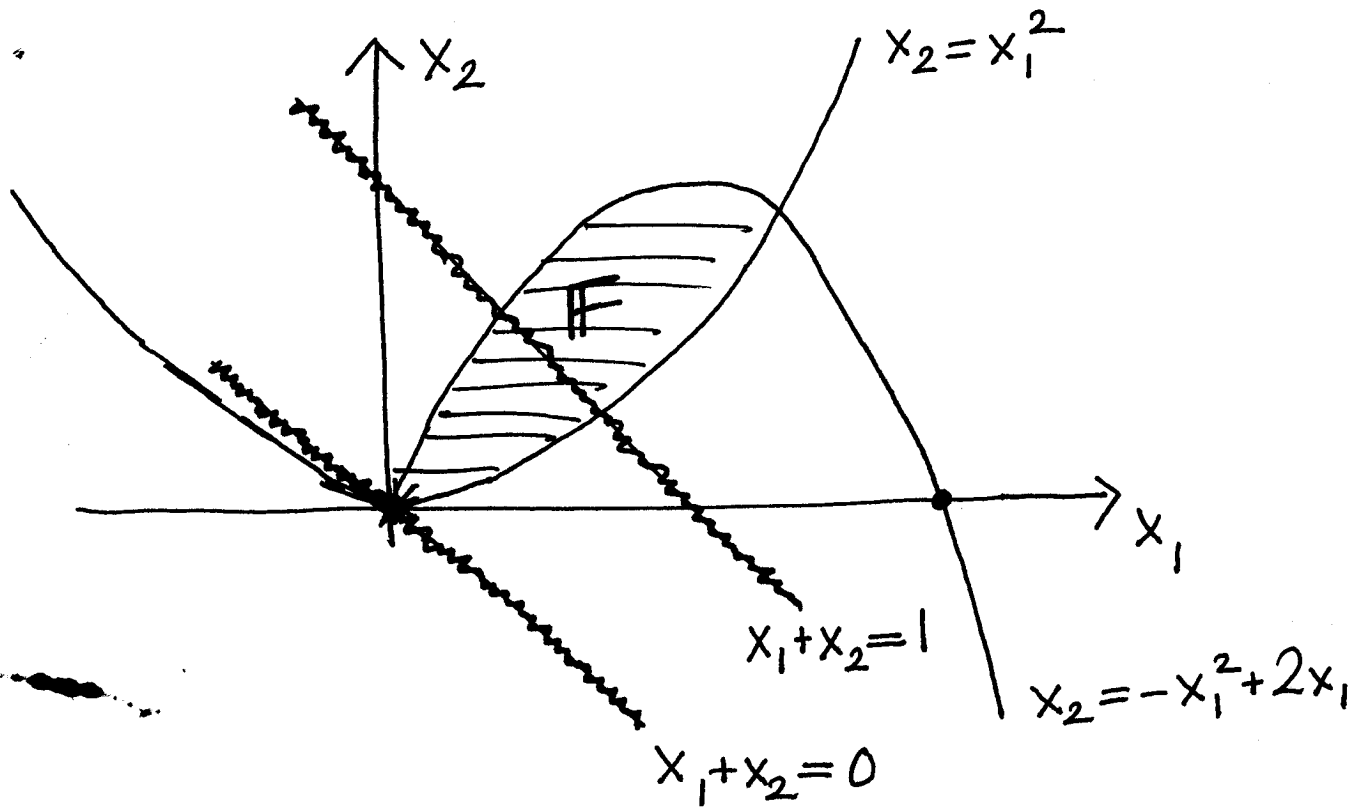
$$(i) \quad c_j(x_*) \geq 0 \quad j=1, \dots, m$$

$$(ii) \quad \nabla f(x_*) = J_a(x_*)^T \lambda \quad \text{for some } \lambda \geq 0$$

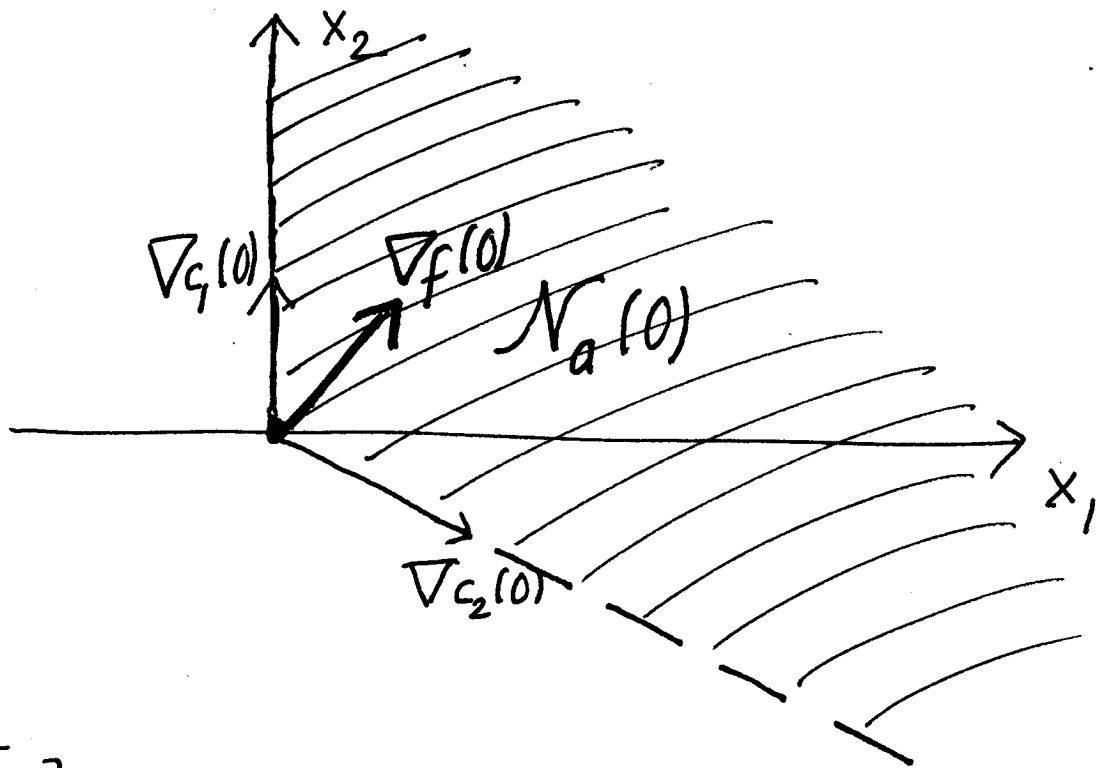
(equivalently $\nabla f(x_*) \in \mathcal{N}_a(x_*)$)

EXAMPLES

$$\begin{aligned} \textcircled{1} \quad & \text{minimize } \underbrace{x_1 + x_2}_{f(x)} \\ & \underbrace{-x_1^2 + x_2}_{c_1(x)} \geq 0 \\ & \underbrace{-x_1^2 + 2x_1 - x_2}_{c_2(x)} \geq 0 \end{aligned}$$



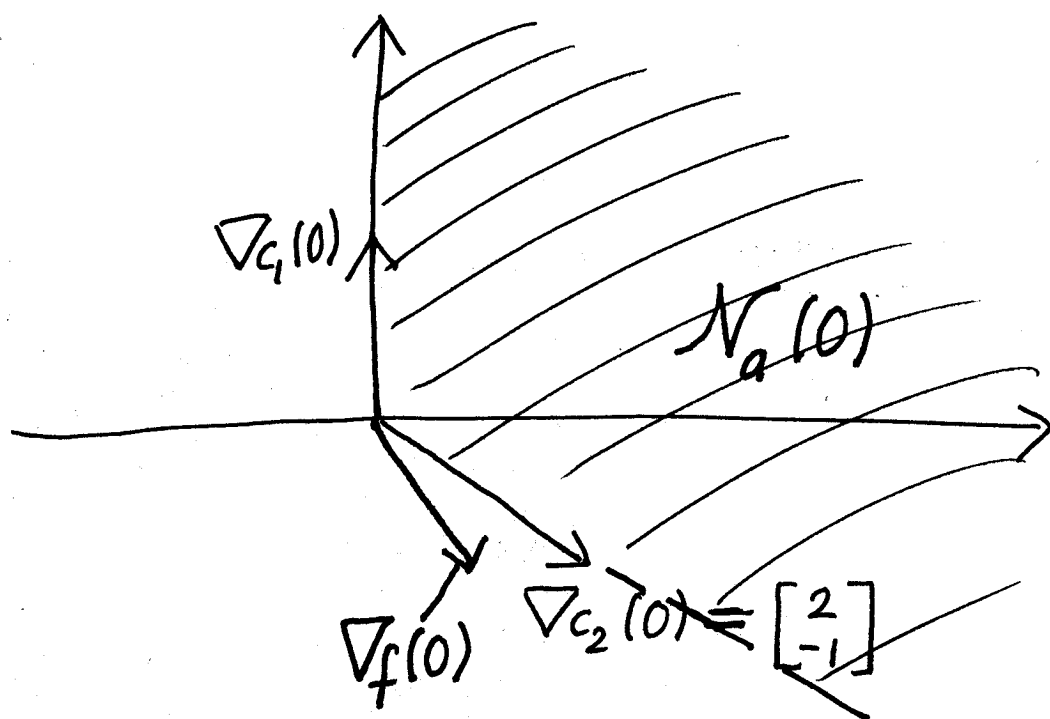
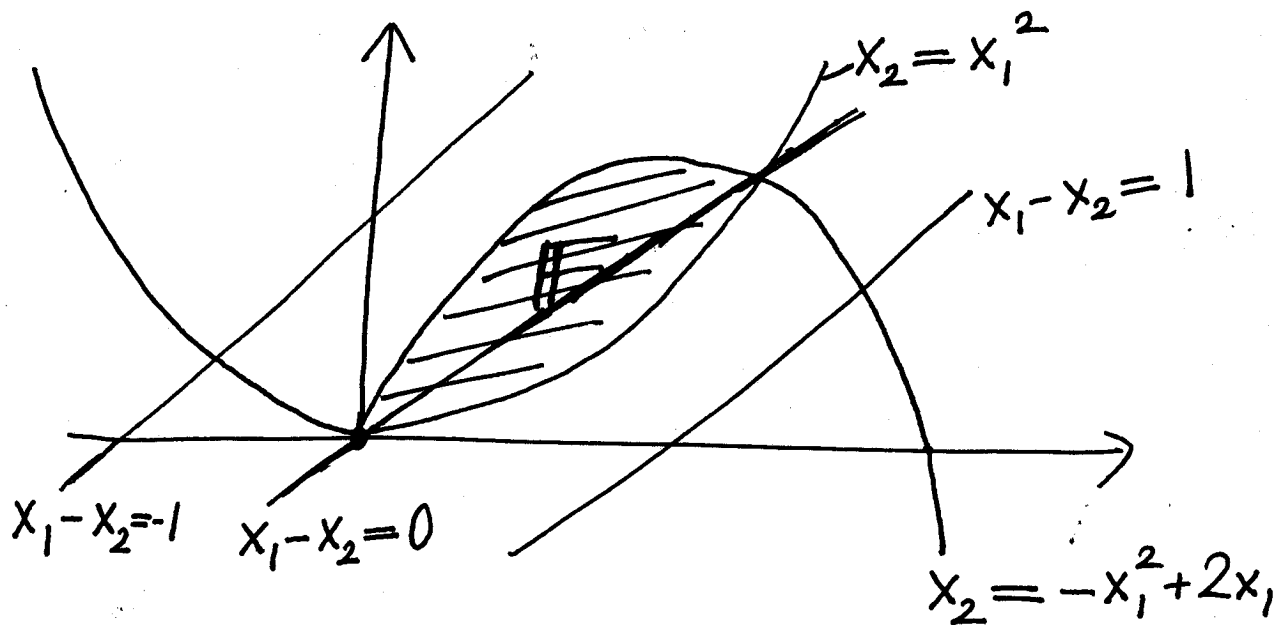
$$\mathcal{N}_a(0) = \left\{ \alpha_0 \begin{bmatrix} 0 \\ 1 \end{bmatrix} + \alpha_1 \begin{bmatrix} 2 \\ -1 \end{bmatrix} : \alpha_0, \alpha_1 \geq 0 \right\}$$



$$\begin{bmatrix} 1 \\ 1 \end{bmatrix} = \nabla f(0) \in \mathcal{N}_a(0)$$

* $(0,0)$ is a local minimizer

② minimize $x_1 - x_2$
 $-x_1^2 + x_2 \geq 0$
 $-x_1^2 + 2x_1 - x_2 \geq 0$



$$\nabla_f(0) = \begin{bmatrix} 1 \\ -1 \end{bmatrix} \notin \mathcal{N}_a(0)$$

* $(0,0)$ is not a local minimizer.

FARKA'S LEMMA

Following conditions are equivalent.

$$(1) \nabla f(x_*)^T p \geq 0 \text{ for all } p \text{ s.t. } J_a(x_*)p \geq 0$$

$$(2) \nabla f(x_*) = J_a(x_*)^T \lambda \text{ for some } \lambda \geq 0$$

PROOF OF (2) \implies (1)

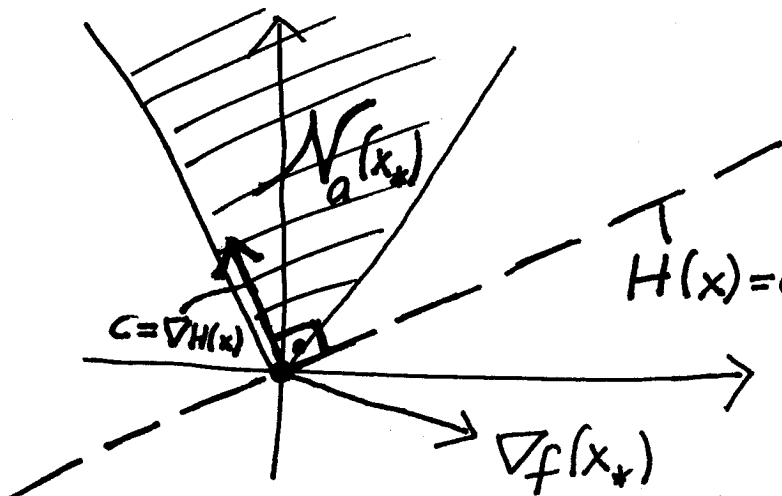
Assume $\nabla f(x_*) = J_a(x_*)^T \lambda$ for some $\lambda \geq 0$

For all p satisfying $J_a(x_*)p \geq 0$

$$\begin{aligned} \nabla f(x_*)^T p &= (J_a(x_*)^T \lambda)^T p \\ &= \underbrace{\lambda^T}_{\geq 0} \underbrace{J_a(x_*)p}_{\geq 0} \geq 0 \end{aligned}$$

JUSTIFICATION FOR (2) \implies (1)

Assume $\nabla f(x_*) \neq J_a(x_*)^T \lambda$ for all $\lambda \geq 0$,
that is $\nabla f(x_*) \notin \mathcal{N}_a(x_*)$.



There exists a separating
hyperplane

$$H(x) = c^T x = 0$$

$$H(x) = c^T x = 0$$

such that

$$(1) \mathcal{N}_a(x_*) \subseteq H_c^+$$

and

$$(2) \nabla f(x_*) \in H_c^- \quad (4)$$

where

$$H_c^+ = \{p : c^T p \geq 0\}$$

$$H_c^- = \{p : c^T p < 0\}$$

Consequently

$$(i) \mathcal{N}_a(x_*) \subseteq H_c^+ \implies \begin{pmatrix} \nabla_{c_{j_1}}(x_*) \in H_c^+ \\ \vdots \\ \nabla_{c_{j_p}}(x_*) \in H_c^+ \end{pmatrix}$$

$$\implies \begin{pmatrix} \nabla_{c_{j_1}}(x_*)^T c \geq 0 \\ \vdots \\ \nabla_{c_{j_p}}(x_*)^T c \geq 0 \end{pmatrix}$$

$$\implies J_a(x_*)c \geq 0$$

$$(ii) \bar{\nabla}_f(x_*) \in H_c^- \implies \bar{\nabla}_f(x_*)^T c < 0$$

There exists a feasible descent direction c such that

$$* \bar{\nabla}_f(x_*)^T c < 0 \quad (c \text{ is descent})$$

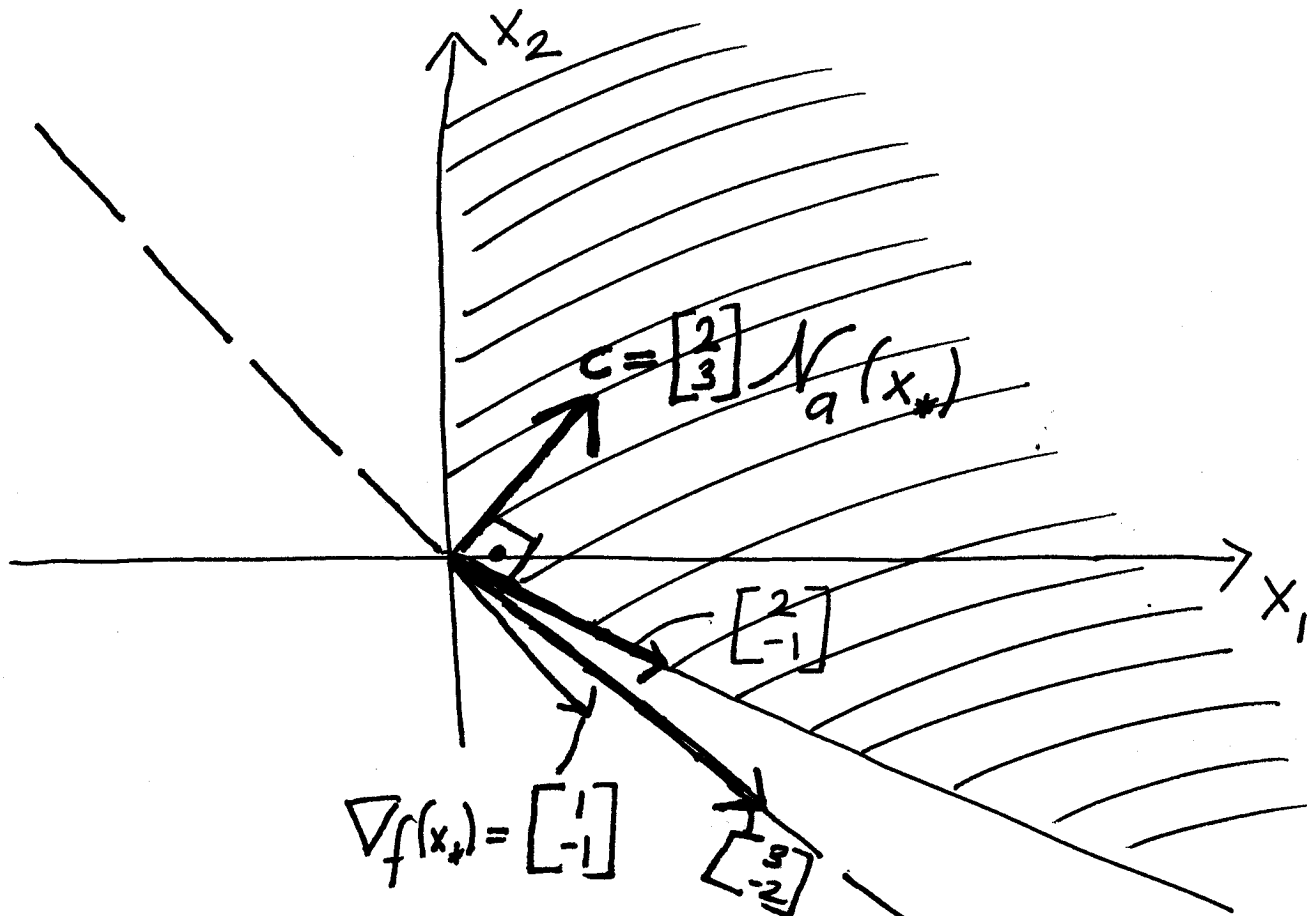
$$* J_a(x_*)c \geq 0 \quad (c \text{ is feasible})$$

EXAMPLE

minimize $x_1 - x_2$

$-x_1^2 + x_2 \geq 0$

$-x_1^2 + 2x_1 - x_2 \geq 0$



Note that $c = \begin{bmatrix} 2 \\ 3 \end{bmatrix}$ is a feasible descent direction

Separating hyperplane

* $\underbrace{\nabla f(0)^T c}_{\text{DESCENT}} = \begin{bmatrix} 1 & -1 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \end{bmatrix} = -1 < 0$ $H(x) = \begin{bmatrix} 2 & 3 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} =$

* $\underbrace{J_a(0) c}_{\text{FEASIBLE}} = \begin{bmatrix} 0 & 1 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} 2 \\ 3 \end{bmatrix} = \begin{bmatrix} 3 \\ 1 \end{bmatrix} \geq 0$

COMPLEMENTARITY FORM

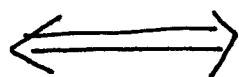
It is more convenient to express

$$\nabla f(x) = J_a(x)^T \lambda = \lambda_1 \nabla c_{j_1}(x) + \dots + \lambda_\ell \nabla c_{j_\ell}(x)$$

in terms of all constraints.

(rather than active constraints.)

$$J_a(x)^T \lambda = \nabla f(x) \text{ for some } \lambda \geq 0$$



$$J(x)^T \tilde{\lambda} = \nabla f(x) \text{ for some } \tilde{\lambda} \geq 0$$

and

$$\boxed{\text{COMPLEMENTARITY CONDITION}} \quad c_j(x) \tilde{\lambda}_j = 0 \quad \left(\begin{array}{l} \text{if } \bar{c}_j(x) \neq 0, \\ \text{then } \tilde{\lambda}_j = 0 \end{array} \right)$$

THM (Optimality Conditions in Complementarity form)

Suppose x_* is a point where constraint qualification holds. If x_* is a local minimizer, then there exists $\lambda \in \mathbb{R}^m$ s.t.

$$(i) \quad c_j(x_*) \geq 0 \quad j=1, \dots, m$$

$$(ii) \quad \nabla f(x_*) = J(x_*)^T \lambda$$

$$(iii) \quad \lambda_j \geq 0 \quad j=1, \dots, m$$

$$(iv) \quad c_j(x_*) \lambda_j = 0 \quad j=1, \dots, m$$

We say that the strict complementarity condition holds at x_* if

$$\begin{aligned} & * \quad c_j(x_*) = 0 \text{ OR } \lambda_j = 0 \\ & \quad \text{BUT NOT } c_j(x_*) = \lambda_j = 0 \\ & \quad \text{for } j = 1, \dots, m \end{aligned}$$

GENERAL NONLINEAR OPTIMIZATION

$$\begin{aligned} \text{(NP)} \quad & \text{minimize} \quad f(x) \\ & x \in \mathbb{R}^n \\ & \text{subject to} \\ & c_j(x) \geq 0 \quad j = 1, \dots, m \\ & \tilde{c}_i(x) = 0 \quad i = 1, \dots, p \end{aligned}$$

NP - Non linear program

(as usual f, c_j, \tilde{c}_i are twice continuously differentiable)

(NP) can be expressed as

$$\begin{aligned} & \text{minimize} \quad f(x) \\ & x \in \mathbb{R}^n \\ & \text{subject to} \\ & c_j(x) \geq 0 \\ & \tilde{c}_i(x) \geq 0 \\ & -\tilde{c}_i(x) \geq 0 \end{aligned}$$

LINEAR INDEPENDENCE CONSTRAINT QUALIFICATION

It can be shown that the constraint qualification holds at x_* if

$\{\nabla c_{j_1}(x_*), \dots, \nabla c_{j_\ell}(x_*), \nabla \tilde{c}_1(x_*), \dots, \nabla \tilde{c}_p(x_*)\}$
is linearly independent where

$$A(x_*) = \{j_1, \dots, j_\ell\}$$

is the set of active inequality constraints.

Assuming constraint qualification holds at $x_* \in \mathbb{F}$.

x_* is a local minimizer

$$\begin{aligned} \nabla f(x_*) &= \sum_{j=1}^m \lambda_j \nabla c_j(x_*) \\ &+ \sum_{i=1}^{\ell} (\lambda_i^+ - \lambda_i^-) \nabla \tilde{c}_i(x_*) \end{aligned}$$

for some $\lambda_j, \lambda_i^+, \lambda_i^- \geq 0$
(and complementarity condition holds)

$$\nabla f(x_*) = \sum_{j=1}^m \lambda_j \nabla c_j(x_*) + \sum_{i=1}^{\ell} \tilde{\lambda}_i \nabla \tilde{c}_i(x_*)$$

for some $\lambda_j \geq 0$ and $\tilde{\lambda}_i$
(and complementarity condition holds) (9)

$$\begin{aligned} & \longleftrightarrow \\ \nabla f(x_*) &= J(x_*)^T \lambda + \tilde{J}(x_*)^T \tilde{\lambda} \end{aligned}$$

where $\lambda \geq 0$. (and complementarity condition holds)

NOTATION
Above and elsewhere

$$J(x) = c'(x) = \begin{bmatrix} \nabla c_1(x)^T \\ \vdots \\ \nabla c_m(x)^T \end{bmatrix}$$

$$\tilde{J}(x) = \tilde{c}'(x) = \begin{bmatrix} \nabla \tilde{c}_1(x)^T \\ \vdots \\ \nabla \tilde{c}_l(x)^T \end{bmatrix}$$

THM ^{-KKT-} (Karush-Kuhn-Tucker conditions for NP)

Let x_* be a point where the constraint qualification holds. If x_* is a local minimizer of (NP), then there exists $\lambda \in \mathbb{R}^m$ and $\tilde{\lambda} \in \mathbb{R}^l$ such that

$$(i) \quad c_j(x_*) \geq 0 \quad j=1, \dots, m$$

$$(ii) \quad \tilde{c}_i(x_*) = 0 \quad i=1, \dots, l$$

$$(iii) \quad \nabla f(x_*) = J(x_*)^T \lambda + \tilde{J}(x_*)^T \tilde{\lambda}$$

$$(iv) \quad \lambda_j \geq 0 \quad j=1, \dots, m$$

$$(v) \quad c_j(x_*) \lambda_j = 0 \quad j=1, \dots, m$$

EXAMPLES

① Linear Program

$$\text{minimize } c^T x$$
$$x \in \mathbb{R}^n$$

subject to

$$Ax = b$$

$$x \geq 0$$

where $A \in \mathbb{R}^{m \times n}$, $b \in \mathbb{R}^m$, $c \in \mathbb{R}^n$.

Constraint qualification always holds.

KKT conditions

If x_* is a local minimizer, there exists $\lambda \in \mathbb{R}^m$ and $M \in \mathbb{R}^m$ such that

(i) $Ax_* = b$

(ii) $x_* \geq 0$

(iii) $c = \underbrace{A^T M}_{\text{equality}} + \underbrace{\lambda}_{\text{inequality}}$

(iv) $\lambda \geq 0$

(v) $\lambda_j(x_*) = 0 \quad j=1, \dots, m$

② Quadratic Program

$$\text{minimize}_{x \in \mathbb{R}^n} \quad \frac{1}{2} x^T H x + g^T x$$

subject to

$$A x \geq b$$

$$\tilde{A} x = \tilde{b}$$

where $A \in \mathbb{R}^{m \times n}$, $\tilde{A} \in \mathbb{R}^{l \times n}$, $b \in \mathbb{R}^m$, $\tilde{b} \in \mathbb{R}^l$

Constraint qualification holds.

KKT conditions

If x_* is a local minimizer, there exist $\lambda \in \mathbb{R}^m$ and $\mu \in \mathbb{R}^l$ such that

$$(i) \quad \tilde{A} x_* = \tilde{b}$$

$$(ii) \quad A x_* \geq b$$

$$(iii) \quad H x_* + g = \tilde{A}^T \mu + A^T \lambda$$

$$(iv) \quad \lambda \geq 0$$

$$(v) \quad \lambda_j (A x_* - b)_j = 0 \quad j=1, \dots, m$$