

LECTURE 19

MATH 409/509

SPRING 2011

NONLINEAR OPTIMIZATION WITH INEQUALITY CONSTRAINTS (PART I)

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

NIP - non linear inequality-constrained program

OBJECTIVE FUNCTION $f: \mathbb{R}^n \rightarrow \mathbb{R}$

CONSTRAINT $c_j: \mathbb{R}^n \rightarrow \mathbb{R}$

are twice continuously differentiable

More difficult than (NEP). An (NEP) can be turned into an (NIP).

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

equivalent to

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

DEFN (Feasible Region)

The set

$$F = \{x \in \mathbb{R}^n : c_j(x) \geq 0 \quad j=1, \dots, m\}$$
 is called the feasible region.

The definitions of a local and a global minimizer are same as the definitions for (NEP) but using the feasible region as defined above.

DEFN (Active & Inactive Constraints)

Let $x_* \in F$. A constraint $c_j(x)$ is said to be

* active at x_* if $c_j(x_*) = 0$, and

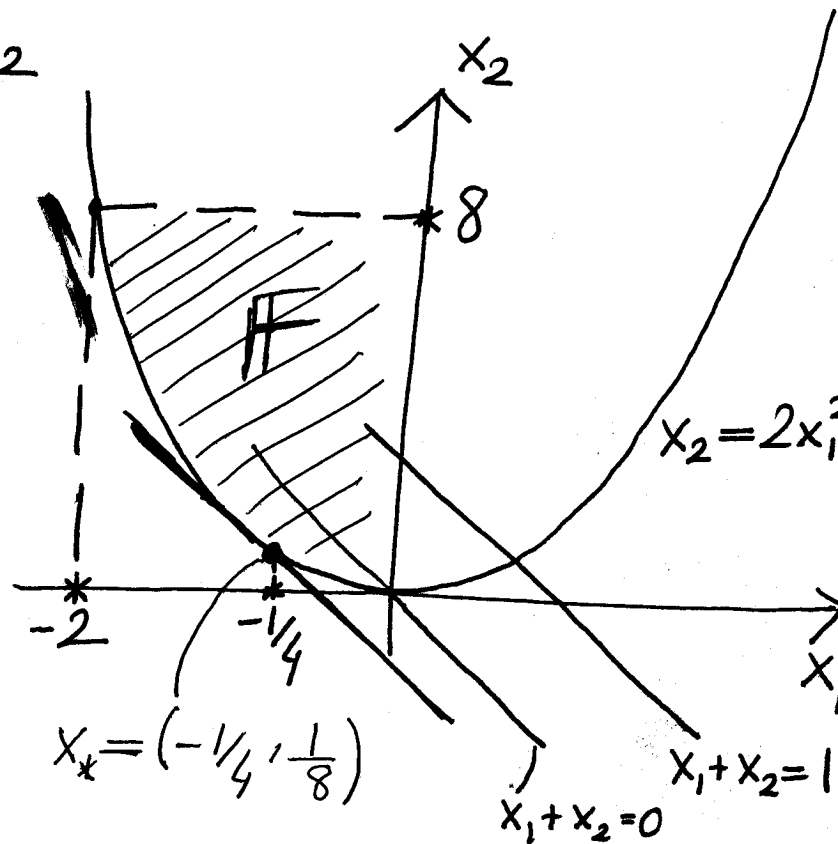
* inactive at x_* if $c_j(x_*) > 0$.

REMARKS

- * The inherent difficulty in (NIP) is that the active constraints at a local minimizer are not known in advance.
- * If the active constraints were known, an (NIP) is simply an (NEP).

EXAMPLE

$$\begin{aligned} & \text{minimize } x_1 + x_2 \\ & x \in \mathbb{R}^2 \\ & \text{subject to} \\ & -2x_1^2 + x_2 \geq 0 \\ & -x_1 \geq 0 \\ & x_1 \geq -2 \\ & -x_2 \geq -8 \end{aligned}$$



At the minimizer $x_* = (-1/4, 1/8)$ only the constraint $-2x_1^2 + x_2 \geq 0$ is active.

The problem is equivalent to

$$\begin{aligned} & \text{minimize } x_1 + x_2 \\ & x \in \mathbb{R}^2 \\ & \text{subject to} \\ & -2x_1^2 + x_2 = 0 \end{aligned}$$

REMARK

In general if we know precisely that constraints j_1, j_2, \dots, j_p are active at a local minimizer, then (NIP) is equivalent to

$$\begin{aligned} & \text{minimize } f(x) \\ & x \in \mathbb{R}^n \\ & \text{subject to} \\ & g_k(x) = 0 \quad k=1, \dots, p \end{aligned}$$

OPTIMALITY CONDITIONS

Suppose $x_* \in F$ is a local minimizer.

For any feasible path $x(\alpha)$

$$\begin{aligned} f(x(\alpha)) &= f(x(0)) + \left. \frac{df(x(\alpha))}{d\alpha} \right|_{\alpha=0} \alpha + \left. \frac{d^2 f(x(\alpha))}{d\alpha^2} \right|_{\alpha=t} \frac{\alpha^2}{2} \\ &= f(x_*) + \alpha \nabla f(x_*)^T x'(0) + c\alpha^2 \end{aligned}$$

where $\alpha > 0$ and $t \in (0, \alpha)$.

Since $x_* \in F$ is a local minimizer

$$0 \leq f(x(\alpha)) - f(x_*) = \alpha (\nabla f(x_*)^T x'(0)) + c\alpha^2$$

\implies

$$\nabla f(x_*)^T x'(0) \geq 0$$

for all feasible path $x(\alpha)$ at x_*

DEFN (Feasible path for (NIP))

A twice continuously differentiable directed parametric curve $x(\alpha): \mathbb{R} \rightarrow \mathbb{R}^n$ is a feasible path at x_* if

(i) $x(0) = x_*$,

(ii) $x'(0) \neq 0$, and

(iii) there exists a $\sigma > 0$ s.t.

$c_j(x(\alpha)) \geq 0$ for all $\alpha \in [0, \sigma)$ and $j=1, \dots, m$ \oplus

DEFN (Tangent Cone)

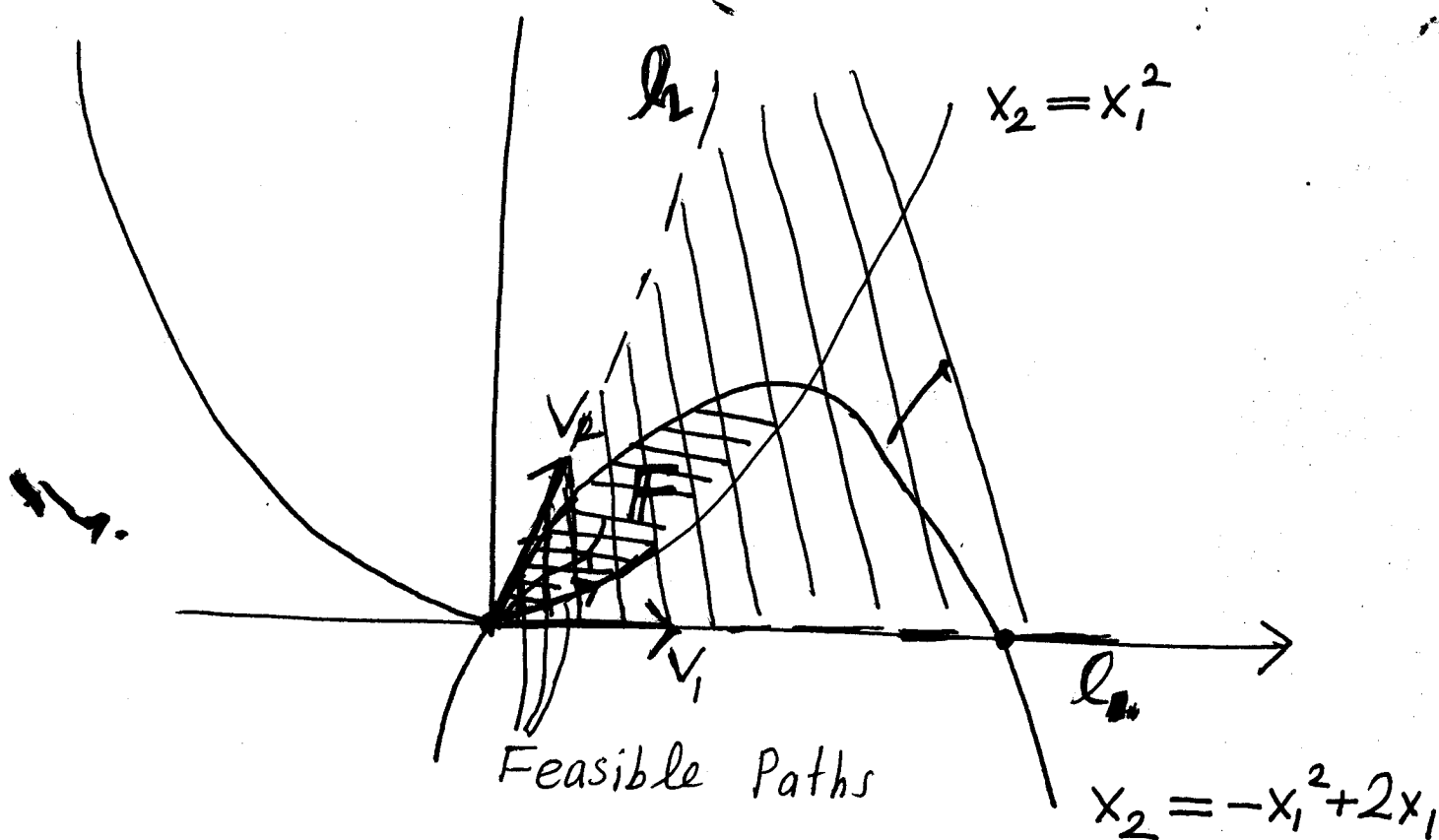
The set

$T^0(x_*) = \{x'(t) : x(t) \text{ is a feasible path}\} \cup \{0\}$
is called the tangent cone at x_* .

EXAMPLE

Consider the constraints

$$(i) -x_1^2 + x_2 \geq 0 \quad (ii) -x_1^2 + 2x_1 - x_2 \geq 0$$



v_1 - vector tangent to $x_2 = x_1^2$
at $(0,0)$

v_2 - vector tangent to $x_2 = -x_1^2 + 2x_1$
at $(0,0)$

$$T^{\circ}(0) = \{ \alpha_1 v_1 + \alpha_2 v_2 : \alpha_1, \alpha_2 \geq 0 \}$$

All linear combinations of v_1, v_2 with positive weights.

l_1 - line tangent to $x_2 = x_1^2$
(and parallel to v_1) at $(0,0)$

l_2 - line tangent to $x_2 = -x_1^2 + 2x_1$
(and parallel to v_2) at $(0,0)$.

$T^{\circ}(0)$ is the region bounded by l_1 and l_2 in the first quadrant.

$$\text{Slope of } l_2 = \left. \frac{d(-x_1^2 + 2x_1)}{dx_1} \right|_{x_1=0} = 2$$

$$\implies v_2 = \alpha_2 \begin{bmatrix} 1 \\ 2 \end{bmatrix} \quad \forall \alpha_2 \in \mathbb{R}^+$$

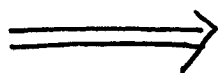
$$\text{Slope of } l_1 = \left. \frac{d(x_1^2)}{dx_1} \right|_{x_1=0} = 0$$

$$\implies v_1 = \alpha_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad \forall \alpha_1 \in \mathbb{R}^+$$

$$T^{\circ}(0) = \left\{ \alpha_1 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \alpha_2 \begin{bmatrix} 1 \\ 2 \end{bmatrix} : \alpha_1, \alpha_2 \geq 0 \right\}$$

CHARAC 1

x_* is a local minimizer



$\nabla f(x_*)^T p \geq 0$ for all $p \in T^{\circ}(x_*)$