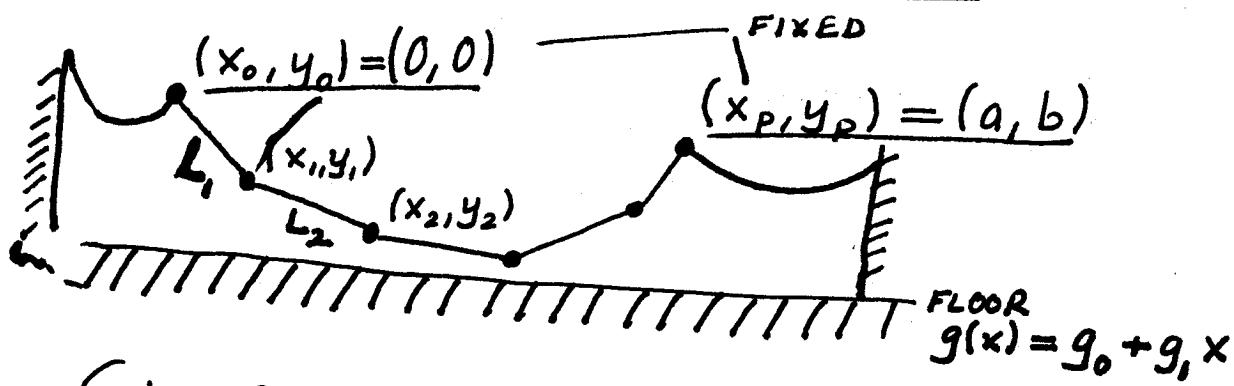


LECTURE 26NONLINEAR PROGRAMS AND THEIR
NUMERICAL SOLUTION

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

Hanging Chain Problem

Given

- * coordinates of the end-joints $(x_0, y_0) = (0, 0)$ and $(x_p, y_p) = (a, b)$
- * length of the j th rod $L_j \quad j=1, \dots, p$
- * all joints must be above the floor, i.e., $g_0 + g_1 x_i \leq y_i$ (1)

PROBLEM

- (1) p rods are connected at joint points with coordinates (x_j, y_j)
 $j = 1, \dots, p$
- (2) The length of j th rod is fixed and equal to L_j . But the coordinates of the j th joint (x_j, y_j) are variables.
- (3) Find the joint coordinates minimizing the total potential

Objective Function

$$\phi(x) = \sum_{j=1}^p L_j \underbrace{\frac{(y_j + y_{j-1})}{2}}_{\text{Potential of the } j\text{th rod}}$$

VECTOR
OF
UNKNOWNs $x = [x_1 \ x_2 \ \dots \ x_p \ y_1 \ y_2 \ \dots \ y_p]^T \in \mathbb{R}^{2p}$

Equality Constraints

$$j=1, \dots, p \quad L_j^2 = (x_j - x_{j-1})^2 + (y_j - y_{j-1})^2$$

that is the length of the j th rod is L_j .

Equivalently

$$\underbrace{(x_j - x_{j-1})^2 + (y_j - y_{j-1})^2}_{\tilde{c}_j(x)} - L_j^2 = 0$$

Inequality Constraints

$$j=0, \dots, p \quad g_0 + g_i x_j \leq y_j$$

that is the j th joint
must be above the floor

~~Objective~~

Equivalently

$$\underbrace{y_i - g_0 - g_i x_i}_{c_i(x)} \geq 0$$

Problem Formulation

$$\underset{x \in \mathbb{R}^{2p+2}}{\text{minimize}} \quad \sum_{j=1}^p L_j \left(\frac{y_j + y_{j-1}}{2} \right)$$

subject to

$$y_i - g_0 - g_i x_i \geq 0 \quad i=0, \dots, p$$

$$(x_j - x_{j-1})^2 + (y_j - y_{j-1})^2 - L_j^2 = 0 \quad j=1, \dots, p$$

(Above in the objective and constraints)
 $x_0 = 0, x_p = a, y_0 = 0, y_p = b$

③

Penalty Function Method

First focus on equality constraints only.

$$\begin{array}{l} \text{minimize } f(x) \\ (\text{NEP}) \quad x \in \mathbb{R}^n \\ \text{subject to} \\ \tilde{c}_j(x) = 0 \quad j = 1, \dots, l \end{array}$$

Basic Idea

- * Replace the problem (NEP) by an unconstrained optimization problem
- * Penalize the violation of equality constraints

Penalty Function

$$P(x, M) = f(x) + \frac{1}{2M} \sum_{j=1}^l \tilde{c}_j^2(x)$$

where $M > 0$ is the penalty parameter.

EXAMPLE

$$\begin{array}{ll} \text{minimize} & x_1 + x_2 \\ x \in \mathbb{R}^2 \\ \text{subject to} \\ x_1^2 + x_2^2 = 1 \end{array}$$

Penalty Function

$$P(x, M) = x_1 + x_2 + \frac{1}{2M} (x_1^2 + x_2^2 - 1)^2$$

Algorithm

$$\begin{array}{ll} \text{minimize} & P(x, M) \\ x \in \mathbb{R}^n \end{array}$$

Solve the unconstrained optimization problem above for various M

- * starting with large M
- * eventually letting $M \rightarrow 0$

EXAMPLE

$$\text{minimize } x_2$$

$$x_1^2 - x_2 = 1$$

Optimal objective function value

$$(x_{*})_2 = (x_{*})_1^2 - 1 = -\underline{\underline{1}}$$

Penalty function

$$P(x, M) = x_2 + \frac{1}{2M} (x_1^2 - x_2 - 1)^2$$

Finding a local min as a function of M

- * $\nabla_x P(x, M) = \begin{bmatrix} 2x_1 (x_1^2 - x_2 - 1)/M \\ 1 - (x_1^2 - x_2 - 1)/M \end{bmatrix}$

- * only stationary point

$$(x_1(M), x_2(M)) = (0, -1-M)$$

- * Hessian

$$\nabla_{xx}^2 P(x, M) = \begin{bmatrix} 0 & 0 \\ 0 & 1/M \end{bmatrix} \geq 0$$

- * Consequently

$$(x_1(M), x_2(M)) = (0, -1-M)$$

is the unique local minimizer.

Letting $M \rightarrow 0$

$$\lim_{M \rightarrow 0} x_2(M) = -1$$

APPROACHES OPTIMAL
OBJECTIVE VALUE

ALGORITHM (Penalty Function)

Given $x_0^s \in \mathbb{R}^n$ and $M_0 > 0$. Let $k=0$.

While $M_k > \epsilon$

- (1) Use Newton's method to find an approximate local minimizer x_k of $P(x, M_k)$ w.r.t. x such that

$$\|\nabla_x P(x_k, M_k)\| < \tau_k.$$

(When choosing the initial guess
 x_k^s for Newton's method use
the previous solution x_{k-1})

- (2) choose a $M_{k+1} \in (0, M_k)$
- (3) $k=k+1$

end

THM (Convergence of Penalty Function Method)

Suppose

$$\lim_{k \rightarrow \infty} M_k = 0 \quad \text{and} \quad \lim_{k \rightarrow \infty} \tau_k = 0.$$

Then

- (i) any limit point x^* of the sequence $\{x_k\}$ / where LICQ holds satisfies the KKT conditions for (NLP) for some Lagrange multiplier λ .

• (ii) Furthermore

$$\lim_{k \rightarrow \infty} -\frac{\tilde{c}_j(x_k)}{M_k} = (\lambda_*)_j.$$

PROOF

Consider

$$\nabla_x P(x_k, M_k) = \nabla_f(x_k) + \frac{1}{2M_k} \sum_{j=1}^l \tilde{c}_j(x_k) \nabla \tilde{c}_j(x_k)$$

Since $\|\nabla_x P(x_k, M_k)\| < \tau_k$ and $\tau_k \rightarrow 0$
as $k \rightarrow \infty$, we have

$$(*) \quad \lim_{k \rightarrow \infty} \left\| \nabla_f(x_k) + \sum_{j=1}^l \frac{\tilde{c}_j(x_k)}{M_k} \nabla \tilde{c}_j(x_k) \right\| = \emptyset$$

=====

$$\lim_{k \rightarrow \infty} \left\| \sum_{j=1}^l \frac{\tilde{c}_j(x_k)}{M_k} \nabla \tilde{c}_j(x_k) \right\| = \lim_{k \rightarrow \infty} \|\nabla_f(x_k)\|$$

=====

$$\left\| \sum_{j=1}^l \tilde{c}_j(x_*) \nabla \tilde{c}_j(x_*) \right\| = \lim_{k \rightarrow \infty} M_k \|\nabla_f(x_k)\| \\ = 0$$

By assumption $\{\nabla \tilde{c}_j(x_*): j=1, \dots, l\}$ is
linearly independent meaning

$$(1) \quad \tilde{c}_j(x_*) = 0 \quad j=1, \dots, l.$$

Define

$$\lambda_k := -\frac{\tilde{c}(x_k)}{M_k},$$

and note that (+) implies

$$\lim_{k \rightarrow \infty} \nabla f(x_k) - J(x_k)^T \lambda_k = 0$$

—————
⇒

$$(2) \quad \nabla f(x_*) = J(x_*) \lambda_*$$

where $\lambda_* = \lim_{k \rightarrow \infty} \lambda_k$. Equations (1) and
(2) show that KKT conditions hold
at x_* for (NEP).

□