### Optimization with constraints

The problem P1:

$$\min f(x)$$
,

subject to

$$x \in C = \{g_i(x) \le 0 \mid i \in 1 : m\}.$$

#### Penalty method

Introduce a penalty function:

$$h(x)$$
  $\begin{cases} = 0 & \text{for } x \in C, \\ > 0 & \text{for } x \in C. \end{cases}$ 

An example of a penalty function

$$h(x) = \sum_{i=1}^{m} (\max(g_i(x), 0))^2.$$

Consider a sequence of unconstrained problems  $\{P_k\}$ :

$$\min_{x \in R^n} (f(x) + A_k h(x))$$

where  $\lim_{k\to\infty} A_k = \infty$ .

Let  $x^k$  be a solution of the problem  $P_k$ . If f and  $g_i$  are continuous functions, the set  $\{x^k\}$  is compact, the solution of the problem P1 exists and subsequence  $\{x^{k_i}\}$  converges, then  $x^* = \lim x^{k_i}$  is a solution of the problem P1.

#### Barrier method

b is a barrier function if

$$b(x) \ge c$$
 for some c if  $x \in C$ ,

$$b(x) \to \infty$$
, if  $g_i(x) \to 0$  for some i.

Examples of barrier functions:

$$b(x) = \sum_{i=1}^{m} (-1/g_i(x)),$$

$$b(x) = \sum_{i=1}^{m} \log(-g_i(x)).$$

Consider a sequence of unconstrained problems  $\{B_k\}$ :

$$\min_{x \in R^n} (f(x) + r_k b(x))$$

where  $r_k > 0$ ,  $\lim_{k \to \infty} r_k = 0$ .

Let  $x^k$  be a solution of the problem  $B_k$ . If f and  $g_i$  are continuous functions, set C is compact and Int(C) is not empty, the solution of the problem P1 exists and subsequence  $\{x^{k_i}\}$  converges, then  $x^* = \lim x^{k_i}$  is a solution of the problem P1.

#### Method of feasible directions

Let  $A(x, \delta) = \{i \in 1 : m : g_i(x) > -\delta\}.$ Let  $x^0 \in C$  be an initial point,  $\delta_0 > 0$ .

# Algorithm.

Let us have  $x^k \in C$  and  $\delta_k > 0$ .

Find a feasible direction p solving the problem L1:

 $\min \eta$ 

subject to

$$egin{aligned} (
abla f(x^k), p) &\leq \eta, \ (
abla g_i(x^k), p) &\leq \eta, \ i \in A(x^k, \delta_k), \ \|p\| &\leq 1. \end{aligned}$$

If  $||p|| = \max_{j} |p_{j}|$ , L1 is a linear programming problem.

Let  $p^k$  and  $\eta_k$  be a solution of L1.

(1) If 
$$\eta_k \ge -\delta_k$$
, set  $x^{k+1} = x^k$ ,  $\delta_{k+1} = \delta_k/2$ .

(2) If  $\eta_k < -\delta_k$ , search for  $\alpha_k$ :

Start with  $\alpha = 1$ . Check the condition:

$$f(x^k + \alpha p^k) \le f(x^k) + \alpha \eta_k / 2,$$
$$g_i(x^k + \alpha p^k) \le 0, \ i \in 1: m.$$

If the conditions are satisfied set  $\alpha_k = \alpha$ , otherwise set  $\alpha := \alpha/2$  until the conditions will be satisfied.

Set 
$$x^{k+1} = x^k + \alpha_k p^k$$
,  $\delta_{k+1} = \delta_k$ .

If f and  $g_i$  are convex functions, the regularity condition is satisfied and set C is compact, then  $f(x^k) \to f(x^*)$ , where  $x^*$  is a solution of the problem P1.

## How to find a feasible solution?

Consider the following problem (P2):

 $\min \eta$ ,

subject to

$$g_i(x) - \eta \le 0, \ i \in 1:m.$$

An initial point  $(x^0, \eta_0)$  for P2 is:  $x^0$  is an arbitrary point in  $\mathbb{R}^n$  and  $\eta_0 = \max_{i \in 1:m} g_i(x_0)$ .

Solve the problem P2 by the method of feasible directions until  $\eta_k \leq 0$ , then  $x^k$  is a feasible solution of P1.

If  $\min \eta > 0$  in P2, then P1 have no feasible solution.