

Unconstrained optimization

Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  be a continuously differentiable function. Consider

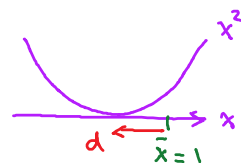
$$(P) \quad \inf_{x \in \mathbb{R}^n} f(x)$$

Recall: If  $n=1$ , then  $\frac{df}{dx} = 0$  is a necessary condition for optimality.

How about general  $n$ ?

Proposition: Let  $f$  be as above,  $\bar{x} \in \mathbb{R}^n$  be arbitrary. If there exists a  $d \in \mathbb{R}^n \setminus \{0\}$  s.t.  $\nabla f(\bar{x})^T d < 0$ , then  $\exists \alpha_0 > 0$  s.t.  $f(\bar{x} + \alpha d) < f(\bar{x})$  for all  $\alpha \in (0, \alpha_0)$ .

Here,  $d$  is called a descent direction of  $f$  at  $\bar{x}$ .



Corollary (First-order necessary optimality condition)  $\nabla f(\bar{x}) = \frac{df}{dx} \Big|_{\bar{x}=1} = 2$

$$d = -1$$

Under the above setting, if  $\bar{x}$  is a local minimum of  $f$ , then  $\nabla f(\bar{x}) = 0$ .

Proof: If  $\nabla f(\bar{x}) \neq 0$ , then take  $d = -\nabla f(\bar{x})$  in the proposition.

Proposition: Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  be a continuous differentiable and convex function. Then,  $\bar{x}$  is global minimizer  $\Leftrightarrow \nabla f(\bar{x}) = 0$ .

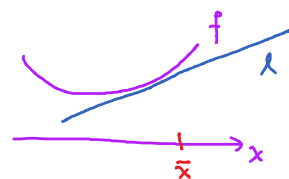
Proof: ( $\Rightarrow$ ) Follow directly from the corollary

( $\Leftarrow$ ) Suppose that  $\nabla f(\bar{x}) = 0$ . Since  $f$  is convex,

$$f(x) \geq \underbrace{f(\bar{x}) + \nabla f(\bar{x})^T (x - \bar{x})}_{\lambda} \quad \forall x$$

$$= f(\bar{x}).$$

This implies that  $\bar{x}$  is a global minimizer.



Proposition (Second-order sufficient optimality condition)

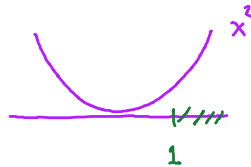
Proposition (Second-order sufficient optimality condition)

Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  be twice continuously differentiable. If  $\nabla f(\bar{x}) = 0$ ,  $\nabla^2 f(\bar{x}) > 0$ , then  $\bar{x}$  is a local minimizer.

Constrained Optimization

Examples:

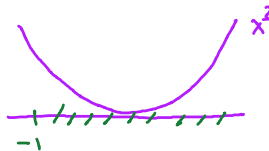
1)  $\inf f(x) = x^2$   
s.t.  $x \geq 1$



$\frac{df}{dx} \Big|_{\bar{x}=1} = 2 \neq 0$

$\frac{df}{dx} = 0 \Rightarrow x=0$   
infeasible

2)  $\inf f(x) = x^2$   
s.t.  $x \geq -1$



$\bar{x} = 0$  optimal

$\frac{df}{dx} \Big|_{\bar{x}=0} = 0$

Let  $f, g_1, \dots, g_{m_1}, h_1, h_2, \dots, h_{m_2}: \mathbb{R}^n \rightarrow \mathbb{R}$  be continuously differentiable functions. Consider

(Q) 
$$\begin{aligned} \inf & f(x) \\ \text{s.t.} & \left. \begin{aligned} g_i(x) &\leq 0, \quad i=1, \dots, m_1, (v_i) \\ h_j(x) &= 0, \quad j=1, \dots, m_2, (w_j) \\ x &\in \mathbb{R}^n \end{aligned} \right\} S: \text{feasible region} \end{aligned}$$

Theorem (Karush-Kuhn-Tucker (KKT) necessary optimality condition)

Let  $\bar{x} \in \mathbb{R}^n$  be a local minimum of (Q). Let

$I = \{ i : g_i(\bar{x}) = 0 \}$

be the active index set. Suppose that  $\{ \nabla g_i(\bar{x}) \}_{i \in I} \cup \{ \nabla h_j(\bar{x}) \}_{j=1}^{m_2}$

are linearly independent. Then,  $\exists$  Lagrange multipliers  $v_1, \dots, v_{m_1};$

$w_1, \dots, w_{m_2} \in \mathbb{R}$  s.t.

(\*) 
$$\nabla f(\bar{x}) + \sum_{i=1}^{m_1} v_i \nabla g_i(\bar{x}) + \sum_{j=1}^{m_2} w_j \nabla h_j(\bar{x}) = 0,$$

$i \notin I \Rightarrow g_i(\bar{x}) < 0$   
 $\Rightarrow v_i = 0$

$$(*) \quad \nabla f(\bar{x}) + \sum_{i=1}^{m_1} v_i \nabla g_i(\bar{x}) + \sum_{j=1}^{m_2} w_j \nabla h_j(\bar{x}) = 0, \quad \Rightarrow v_i = 0$$

$$(**) \quad v_i \geq 0, \quad \underbrace{v_i g_i(\bar{x}) = 0}_{\text{complementarity}} \quad \forall i.$$

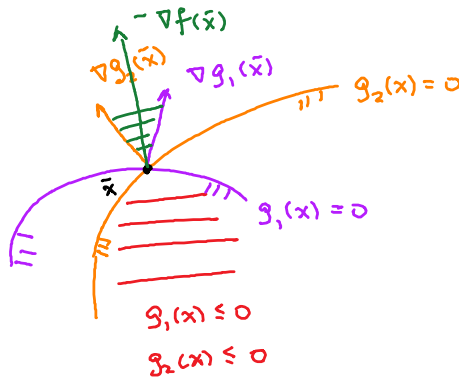
Illustration: For simplicity, suppose that there is no equality constraint.

Consider

$$m_1 = 2,$$

$$m_2 = 0$$

$$I = \{1, 2\}$$



From (\*), (\*\*),

$$-\nabla f(\bar{x}) = v_1 \nabla g_1(\bar{x}) + v_2 \nabla g_2(\bar{x})$$

$$v_1, v_2 \geq 0$$

Idea:

- 1) If  $\bar{x}$  is a local minimum, then  $\nexists d$  s.t.  $d$  is simultaneously a descent direction and a feasible direction.
- 2) To stay feasible, intuitively we need to decrease the value of  $g_1, g_2$ ; e.g., find  $d \in \mathbb{R}^n$  s.t.  $\nabla g_i(\bar{x})^T d < 0, i=1, 2$ .

But for such  $d$ , from (\*),

$$-\nabla f(\bar{x})^T d = \underbrace{v_1}_{\geq 0} \underbrace{\nabla g_1(\bar{x})^T d}_{< 0} + \underbrace{v_2}_{\geq 0} \underbrace{\nabla g_2(\bar{x})^T d}_{< 0} < 0$$

↑ assume  $(v_1, v_2) \neq 0$

$\Rightarrow \nabla f(\bar{x})^T d > 0 \Rightarrow d$  is not a descent direction of  $f$ .