

# 1 Constrained Optimization (May 26)

Consider the following minimization problem:

$$\begin{aligned} &\text{minimize } f(x, y) = xy \\ &\text{subject to } x^2 + y^2 \leq 1 \end{aligned}$$

Let  $\Omega$  be the feasible set. The feasible directions at a point  $(x_0, y_0) \in \Omega$  are the  $(v, w) \in \mathbb{R}^2$  such that  $(v, w) \cdot (x_0, y_0) < 0$ , or  $vx_0 + wy_0 < 0$ . By the FONC for a minimizer,  $\nabla f(x_0, y_0) \cdot (v, w) \geq 0$ , so  $wx_0 + vy_0 \geq 0$ . Note that a local minimum must occur on the boundary. (Why?) We have three cases, depending on the sign of  $x_0 + y_0$ .

- (i)  $x_0 + y_0 < 0$ : can't occur
- (ii)  $x_0 + y_0 > 0$ : can't occur
- (iii)  $x_0 + y_0 = 0$ : good!

(This part could not be finished as attention had to be diverted from the lecture.)

## 1.1 Second Order Necessary Condition for a Local Minimizer

**Theorem 1.1.** (Second order sufficient condition for a local minimizer) Let  $f$  be  $C^2$  on  $\Omega \subseteq \mathbb{R}^n$  and suppose  $x_0 \in \Omega$  satisfies

- (i)  $\nabla f(x_0) \cdot v \geq 0$  for all feasible directions  $v$  at  $x_0$ ,
- (ii) if  $\nabla f(x_0) \cdot v = 0$  for some such  $v$ , then  $v^T \nabla^2 f(x_0) v > 0$ .

Then  $x_0$  is a local minimizer of  $f$  on  $\Omega$ .

## 1.2 Optimization with Equality Constraints

Consider the minimization problem

$$\begin{aligned} &\text{minimize } f(x, y) \\ &\text{subject to } h(x, y) = x^2 + y^2 - 1 = 0 \end{aligned}$$

Suppose  $(x_0, y_0)$  is a local minimizer. Two cases:

1.  $\nabla f(x_0, y_0) \neq 0$ : we claim that  $\nabla f(x_0, y_0)$  is perpendicular to the tangent space to the unit circle  $h^{-1}(\{0\})$  at  $(x_0, y_0)$ . If this is not the case, then we obtain a contradiction by looking at the level sets of  $f$ , to which  $\nabla f$  is perpendicular. Therefore  $\nabla f(x_0, y_0) = \lambda \nabla h(x_0, y_0)$  for some  $\lambda$ .
2.  $\nabla f(x_0, y_0) = 0$ : as in the previous case,  $\lambda = 0$ .

In either case, at a local minimizer, the gradient of the function to be minimized is parallel to the gradient of the constraints.

We now recall some elementary differential geometry.

**Definition 1.** For us, a surface is the set of common zeroes of a finite set of  $C^1$  functions.

**Definition 2.** For us, a differentiable curve on the surface  $M \subseteq \mathbb{R}^n$  is the image of a  $C^1$  function  $x : (a, b) \rightarrow M$ .

**Definition 3.** Let  $x(s)$  be a differentiable curve on  $M$  that passes through  $x_0 \in M$  at time  $x(0) = x_0$ . The velocity vector  $v = \frac{d}{ds}\big|_{s=0} x(s)$  of  $x(s)$  at  $x_0$  is, for us, said to be a tangent vector to the surface  $M$  at  $x_0$ . The set of all tangent vectors to  $M$  at  $x_0$  is called the tangent space to  $M$  at  $x_0$  and is denoted by  $T_{x_0}M$ .

**Definition 4.** Let  $M = \{x \in \mathbb{R}^n : h_1(x) = \cdots = h_k(x) = 0\}$  be a surface. If  $\nabla h_1(x_0), \dots, \nabla h_k(x_0)$  are all linearly independent, then  $x_0$  is said to be a regular point of  $M$ .

**Theorem 1.2.** At a regular point  $x_0 \in M$ , the tangent space  $T_{x_0}M$  is given by

$$T_{x_0}M = \{y \in \mathbb{R}^n : \nabla \mathbf{h}(x_0)y = 0\}.$$

*Proof.* It's in the book. Use the implicit function theorem. □

**Lemma 1.3.** Let  $f, h_1, \dots, h_k$  be  $C^1$  functions on the open set  $\Omega \subseteq \mathbb{R}^n$ . Let  $x_0 \in M = \{x \in \Omega : h_1(x) = \cdots = h_k(x) = 0\}$ . Suppose  $x_0$  is a local minimizer of  $f$  subject to the constraints  $h_i(x) = 0$ . Then  $\nabla f(x_0)$  is perpendicular to  $T_{x_0}M$ .

*Proof.* Without loss of generality, suppose  $\Omega = \mathbb{R}^n$ . Let  $v \in T_{x_0}M$ . Then  $v = \frac{d}{ds}\big|_{s=0} x(s)$  for some differentiable curve  $x(s)$  in  $M$  with  $x(0) = x_0$ . Since  $x_0$  is a local minimizer of  $f$ ,  $0$  is a local minimizer of  $f \circ x$ , so  $\nabla f(x_0) \cdot x'(0) = \nabla f(x_0) \cdot v = 0$ . □