

Ch.5 Elements of Calculus**5.1 Sequences and Limits**

- A number $x^* \in \mathfrak{R}$ is called the limit of the sequence $\{x_k\}$ if for any positive ϵ there is a number such that for all $k > K$, $|x_k - x^*| < \epsilon$

$$x^* = \lim_{k \rightarrow \infty} x_k$$

- A sequence that has a limit is called a convergent sequence.
- **Theorem 5.1** *A convergent sequence has only one limit*
- A sequence $\{x^{(k)}\}$ in \mathfrak{R}^n is bounded if there exists a number $B \geq 0$ such that $\|x^{(k)}\| \leq B$ for all $k = 1, 2, \dots$.
- **Theorem 5.2** *Every convergent sequence is bounded*
- For a sequence $\{x_k\}$ in \mathfrak{R} , a number B is called an upper bound (lower bound) if $x_k \leq B$ ($x_k \geq B$) for all $k = 1, 2, \dots$.
- Any sequence $\{x_k\}$ in \mathfrak{R} that has an upper bound has a least upper bound (also called the supremum), which is the smallest number B that is an upper bound of $\{x_k\}$.
- **Lemma 5.1** *Let $A \in \mathfrak{R}^{n \times n}$. Then, $\lim_{k \rightarrow \infty} A^k = O$ if and only if the eigenvalues of A satisfy $|\lambda_i(A)| < 1$, $i = 1, 2, \dots, n$.*

- **Lemma 5.2** *The series of $n \times n$ matrices*

$$I_n + A + A^2 + \cdots + A^k + \cdots$$

converges if and only if $\lim_{k \rightarrow \infty} A^k = O$. In this case the sum of the series equals $(I_n - A)^{-1}$.

5.2 Differentiability

- A function $A : \mathbb{R}^n \rightarrow \mathbb{R}^m$ is affine if there exists a linear function $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$ and a vector $y \in \mathbb{R}^m$ such that

$$A(x) = L(x) + y$$

for every $x \in \mathbb{R}^n$.

- A function $f : \Omega \rightarrow \mathbb{R}^m$, $\Omega \subset \mathbb{R}^n$, is said to be differentiable at $x_o \in \Omega$ if there is an affine function that approximates f near x_o , that is, there exists a linear function $L : \mathbb{R}^n \rightarrow \mathbb{R}^m$ such that

$$\lim_{x \rightarrow x_o, x \in \Omega} \frac{\|f(x) - (L(x - x_o) + f(x_o))\|}{\|x - x_o\|} = 0$$

5.3 The Derivative Matrix

- The matrix L is called the Jacobian matrix, or derivative matrix, of f at x_0 , and is denoted $Df(x_0)$.
- Given $f : \mathbb{R}^n \rightarrow \mathbb{R}$, if ∇f is differentiable, we say that f

is twice differentiable, and the derivative of ∇f as

$$D^2 f = \begin{bmatrix} \frac{\partial^2 f}{\partial x_1^2} & \frac{\partial^2 f}{\partial x_2 \partial x_1} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_1} \\ \frac{\partial^2 f}{\partial x_1 \partial x_2} & \frac{\partial^2 f}{\partial x_2^2} & \cdots & \frac{\partial^2 f}{\partial x_n \partial x_2} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial^2 f}{\partial x_1 \partial x_n} & \frac{\partial^2 f}{\partial x_2 \partial x_n} & \cdots & \frac{\partial^2 f}{\partial x_n^2} \end{bmatrix}$$

- The matrix $D^2 f(x)$ is called the Hessian matrix of f at x , and is often also denoted $F(x)$.
- A function $f : \Omega \rightarrow \mathbb{R}^m$, $\Omega \subset \mathbb{R}^n$, is said to be continuously differentiable on Ω if it is differentiable, and $Df : \Omega \rightarrow \mathbb{R}^{m \times n}$ is continuous. In this case, we write $f \in C^1$.
- If the components of f have continuous partial derivatives of order p , then we write $f \in C^p$.
- Hessian matrix of a function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ at x is symmetric if f is twice continuously differentiable at x .

5.4 Differentiation Rules

- **Theorem 5.6 (Chain Rule)** *Let $g : D \rightarrow \mathbb{R}$ be differentiable on an open set $D \subset \mathbb{R}^n$, and let $f : (a, b) \rightarrow D$ be differentiable on (a, b) . Then, the composite function $h : (a, b) \rightarrow \mathbb{R}$ given by*

$h(t) = g(f(t))$ is differentiable on (a, b) , and

$$h'(t) = Dg(f(t))Df(t) = \nabla g(f(t))^T \begin{bmatrix} f'_1(t) \\ \vdots \\ f'_n(t) \end{bmatrix}$$

5.5 Level Sets and Gradients

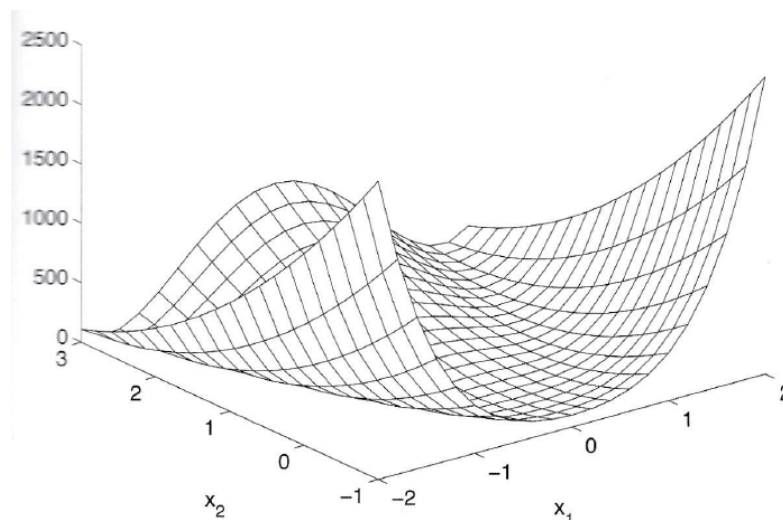
- The level set of a function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ at level c is the set of points

$$S = \{x : f(x) = c\}$$

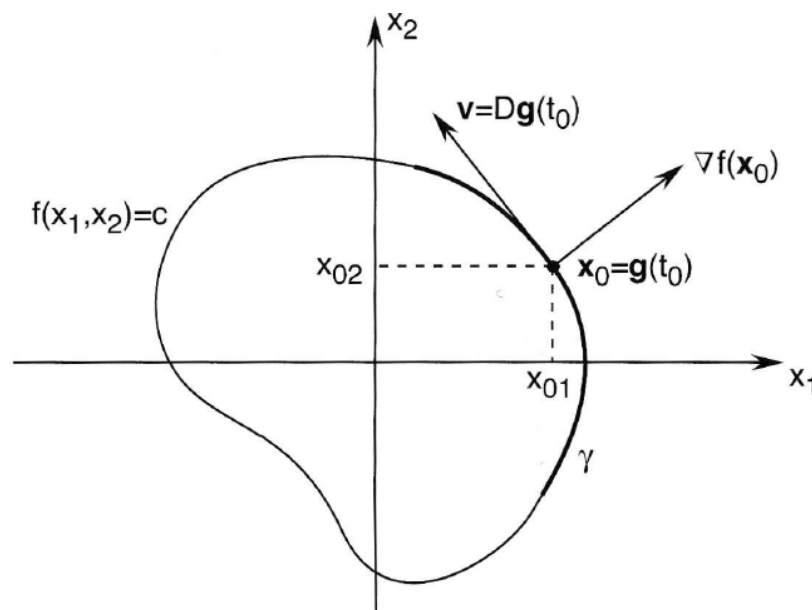
- Example 5.2 consider the following real-valued function on \mathbb{R}^2 :

$$f(x) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2, \quad x = [x_1, x_2]^T$$

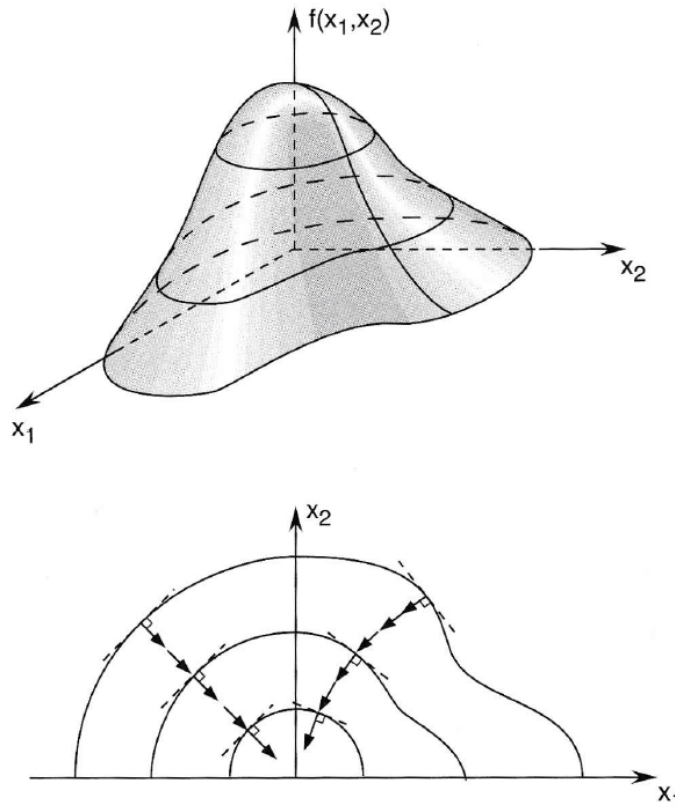
♣ Fig. 5.2 on page 61.



- **Theorem 5.7** *The vector $\nabla f(x_0)$ is orthogonal to the tangent vector to an arbitrary smooth curve passing through x_0 on the level set determined by $f(x) = f(x_0)$.*
- ♣ Fig. 5.4 on page 62.



- $\nabla f(x_0)$ is the direction of maximum rate of increase of f at x_0 .
- ♣ Fig. 5.5 on page 63.



5.6 Taylor Series

- Theorem 5.8 Taylor's theorem** Assume that a function $f : \mathbb{R} \rightarrow \mathbb{R}$ is m times continuously differentiable on an interval $[a, b]$. Denote $h = b - a$. Then

$$\begin{aligned}
 f(b) &= f(a) + \frac{h}{1!} f^{(1)}(a) + \frac{h^2}{2!} f^{(2)}(a) + \cdots \\
 &\quad + \frac{h^{(m-1)}}{(m-1)!} f^{(m-1)}(a) + R_m
 \end{aligned}$$

where $f^{(i)}$ is the i th derivative of f , and

$$R_m = \frac{h^m(1-\theta)^{m-1}}{(m-1)!} f^{(m)}(a+\theta h) = \frac{h^m}{m!} f^{(m)}(a+\theta' h)$$

with $\theta, \theta' \in (0, 1)$.

- Taylor series expansion of a real-valued function $f : \mathbb{R}^n \rightarrow \mathbb{R}$ about the point $x_0 \in \mathbb{R}^n$.

$$\begin{aligned} f(x) &= f(x_0) + \frac{1}{1!} Df(x_0)(x - x_0) \\ &\quad + \frac{1}{2!} (x - x_0)^T D^2 f(x_0)(x - x_0) + o(\|x - x_0\|^2) \end{aligned}$$