

微積分第10回講義ノート

多変数の微分
合成関数の微分
2階の微分
 $f: \mathbb{R}^2 \rightarrow \mathbb{R}$
 $f(x, y)$

2重線型関数

$f(x, y): \mathbb{R}^2 \rightarrow \mathbb{R}$ の線型写像
1×2

$$\left(\frac{\partial f}{\partial x}(x, y), \frac{\partial f}{\partial y}(x, y) \right)$$

$f': \mathbb{R}^2 \rightarrow L(\mathbb{R}^2; \mathbb{R})$ — 本質的には \mathbb{R}^2 と同じ
 $\mathbb{R}^2 \rightarrow \mathbb{R}$ の線型写像の全体

$$f''(x, y) \quad 1 \times 2$$

$f''(x, y): \mathbb{R}^2 \rightarrow L(\mathbb{R}^2; \mathbb{R})$ の線型写像

$$f''(x, y)(x)(y) \in \mathbb{R}$$

$$(P_{ij}) \left(\begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \begin{pmatrix} y_1 \\ y_2 \end{pmatrix} \right) \quad \begin{matrix} i=1,2 \\ j=1,2 \end{matrix}$$

$$= P_{11}x_1y_1 + P_{12}x_1y_2 + P_{21}x_2y_1 + P_{22}x_2y_2$$

$$P_{11} = \frac{\partial}{\partial x} \frac{\partial f}{\partial x} = \frac{\partial^2 f}{\partial x^2}$$

$$P_{12} = \frac{\partial}{\partial y} \frac{\partial f}{\partial x} = \frac{\partial^2 f}{\partial y \partial x}$$

$$P_{22} = \frac{\partial}{\partial y} \frac{\partial f}{\partial y} = \frac{\partial^2 f}{\partial y^2}$$

$$P_{21} = \frac{\partial}{\partial x} \frac{\partial f}{\partial y} = \frac{\partial^2 f}{\partial x \partial y}$$

* 空間 (数学用語): 物の集まり

$$L(\mathbb{R}^2; L(\mathbb{R}^2; \mathbb{R}))$$

$$= L(\mathbb{R}^2; \mathbb{R}^2; \mathbb{R})$$

$$\mathbb{R}^2 \times \mathbb{R}^2 \rightarrow \mathbb{R} \quad \underline{\text{2重線型写像}}$$

$$f(x, y) = x^3 + 2xy + y^3$$

$$\frac{\partial f}{\partial x} = 3x^2 + 2y \quad \frac{\partial f}{\partial y} = 2x + 3y^2$$

$$\frac{\partial}{\partial y} \frac{\partial f}{\partial x} = 2 = \frac{\partial}{\partial x} \frac{\partial f}{\partial y} = 2$$

偏微分は順序によらない

$$f: \mathbb{R}^2 \rightarrow \mathbb{R}$$

$$P_{12} = P_{21}$$

2重線型写像は対称

$$x \cdot y = y \cdot x$$

$$|x \cdot y| = -|y \cdot x|$$

$$\begin{matrix} P_{11} & P_{12} \\ \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \\ P_{21} & P_{22} \end{matrix}$$

$f: \mathbb{R}^n \rightarrow \mathbb{R}^m$ $x, a \in \mathbb{R}^n$
 $f'(x)(a)$
 $f(x+ad_1) - f(x) = \overset{\mathbb{R}^m}{f'(x)(a)} d_1 \quad (\forall d_1 \in D)$ $f': \mathbb{R}^n \rightarrow L(\mathbb{R}^n; \mathbb{R}^m)$

$f'(x+bd_2) - f'(x) = f''(x)(b) d_2 \quad (\forall d_2 \in D)$

$f'(x+bd_2)(a) d_1 - f'(x)(a) d_1 = f''(x)(b)(a) d_2 d_1$

$f(x+bd_2+ad_1) - f(x+bd_2) = f(x+ad_1) - f(x)$

(左辺) = $f(x+ad_1+bd_2) - f(x+ad_1) - f(x+bd_2) + f(x)$

$f(x+ad_1+bd_2) - f(x+ad_1) = f'(x+ad_1)(b) d_2$
 $f(x+bd_2) - f(x) = f'(x)(b) d_2$
 $f'(x+ad_1)(b) d_2 - f'(x)(b) d_2 = \underline{f''(x)(a)(b) d_1 d_2}$

極値 $f: \mathbb{R} \rightarrow \mathbb{R}$ $f(x) = x^3$
 $f(x, y)$ 必要条件 $f'(x) = 3x^2$
 $f'(x) = 0$ $x = 0$ 極値ではない

増減表 〈十分条件〉
 $f(x, y)$ $f'(x) = 0$
 $\frac{\partial f}{\partial x} = \frac{\partial f}{\partial y} = 0$ $f''(x) > 0$ (極小値)
 $f''(x) < 0$ (極大値)
 $f''(x) = 0$ (極値なし)

$x \in \mathbb{R}^2$ 任意の0でないベクトル $a \in \mathbb{R}^2$ について
 $f(x)$ が x で 極大値
 $\Leftrightarrow t \in \mathbb{R} \rightarrow f(x+at)$ が $t=0$ で 極大値

$a = e$ $g'(a)$
 $g''(a)$