

M3/M4S3 STATISTICAL THEORY II

TWO USEFUL RESULTS

Notation: First Derivatives

Suppose $f : \mathbb{R}^k \rightarrow \mathbb{R}$ is a real function, and denote the $(1 \times k)$ vector of *first* partial derivatives by $\dot{\mathbf{f}}$, that is, for $\mathbf{x} \in \mathbb{R}^k$,

$$\dot{\mathbf{f}}(\mathbf{x}) = \left[\frac{\partial f(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_k} \right]$$

By extension, if $\mathbf{f} : \mathbb{R}^k \rightarrow \mathbb{R}^d$ is a real function, then we regard it as a $d \times 1$ vector

$$\mathbf{f}(\mathbf{x}) = \begin{bmatrix} f_1(\mathbf{x}) \\ \vdots \\ f_d(\mathbf{x}) \end{bmatrix}$$

and define the $(d \times k)$ matrix of first partial derivatives with $(j, l)^{\text{th}}$ element

$$\frac{\partial f_j(\mathbf{x})}{\partial x_l},$$

that is,

$$\dot{\mathbf{f}}(\mathbf{x}) = \begin{bmatrix} \frac{\partial f_1(\mathbf{x})}{\partial x_1} & \dots & \frac{\partial f_1(\mathbf{x})}{\partial x_k} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_d(\mathbf{x})}{\partial x_1} & \dots & \frac{\partial f_d(\mathbf{x})}{\partial x_k} \end{bmatrix}$$

Second Derivatives

Suppose $f : \mathbb{R}^k \rightarrow \mathbb{R}$ is a real function, we define the $(k \times k)$ matrix of *second* partial derivatives, $\ddot{f}(\mathbf{x})$, with $(j, l)^{\text{th}}$ element

$$\frac{\partial^2}{\partial x_j \partial x_l} f(\mathbf{x}),$$

that is,

$$\ddot{f}(\mathbf{x}) = \begin{bmatrix} \frac{\partial^2 f(\mathbf{x})}{\partial x_1^2} & \dots & \frac{\partial^2 f(\mathbf{x})}{\partial x_1 \partial x_k} \\ \vdots & \ddots & \vdots \\ \frac{\partial^2 f(\mathbf{x})}{\partial x_k \partial x_1} & \dots & \frac{\partial^2 f(\mathbf{x})}{\partial x_k^2} \end{bmatrix}$$

The Mean-Value Theorem

Suppose that $\mathbf{f} : \mathbb{R}^k \rightarrow \mathbb{R}^d$ is a real function, and that $\dot{\mathbf{f}}(\mathbf{x})$ is continuous in the ball of radius $r > 0$ centered at \mathbf{x}_0 , that is, in

$$\{\mathbf{x} : \|\mathbf{x} - \mathbf{x}_0\| < r\}.$$

Then for $\|\mathbf{t}\| < r$, $\mathbf{t} \in \mathbb{R}^k$,

$$\mathbf{f}(\mathbf{x}_0 + \mathbf{t}) = \mathbf{f}(\mathbf{x}_0) + \left\{ \int_0^1 \dot{\mathbf{f}}(\mathbf{x}_0 + u\mathbf{t}) \, du \right\} \mathbf{t}$$

Proof. Let $\mathbf{h}(u) = \mathbf{f}(\mathbf{x}_0 + u\mathbf{t})$, so that, by the chain rule, $\dot{\mathbf{h}}(u) = \dot{\mathbf{f}}(\mathbf{x}_0 + u\mathbf{t})\mathbf{t}$. Then

$$\left\{ \int_0^1 \dot{\mathbf{f}}(\mathbf{x}_0 + u\mathbf{t}) \, du \right\} \mathbf{t} = \int_0^1 \dot{\mathbf{h}}(u) \, du = \mathbf{h}(1) - \mathbf{h}(0) = \mathbf{f}(\mathbf{x}_0 + \mathbf{t}) - \mathbf{f}(\mathbf{x}_0).$$

and the result follows.

Taylor's Theorem

Suppose that $f : \mathbb{R}^k \rightarrow \mathbb{R}$ is a real function, and that $\ddot{f}(\mathbf{x})$ is continuous in the ball of radius $r > 0$ centered at \mathbf{x}_0 , that is, in

$$\{\mathbf{x} : \|\mathbf{x} - \mathbf{x}_0\| < r\}.$$

Then for $\|\mathbf{t}\| < r$, $\mathbf{t} \in \mathbb{R}^k$,

$$f(\mathbf{x}_0 + \mathbf{t}) = f(\mathbf{x}_0) + \dot{f}(\mathbf{x}_0)\mathbf{t} + \mathbf{t}^\top \left\{ \int_0^1 \int_0^1 v \ddot{f}(\mathbf{x}_0 + uv\mathbf{t}) \, dudv \right\} \mathbf{t}$$

Note that in these two results,

$$\int_0^1 \dot{\mathbf{f}}(\mathbf{x}_0 + u\mathbf{t}) \, du$$

is a $d \times k$ matrix, and

$$\int_0^1 \int_0^1 v \ddot{f}(\mathbf{x}_0 + uv\mathbf{t}) \, dudv$$

is a $k \times k$ matrix.