

if f has a local extremum at a point $p \in \mathbb{R}^n$, then p is a critical point of f . For, if f has a local extremum at $p = (p_1, p_2, \dots, p_n)$, then for any $i = 1, 2, \dots, n$ the function f_i defined by $f_i(t) = f(p_1, p_2, \dots, p_{i-1}, t, p_{i+1}, \dots, p_n)$ has a local extremum at $t = p_i$. So, by an elementary single variable argument

$$\frac{\partial f(p)}{\partial t_i} = \frac{d f_i(p_i)}{dt} = 0$$

Thus p is a critical point of f . But critical points are not necessarily local extrema.

The second order partial derivatives of f at a critical point p can be used to test a local extremum at p . These partial derivatives determine a matrix $A(p)$ in which the row i column j entry is

$$\frac{\partial^2 f(p)}{(\partial t_i)(\partial t_j)}$$

This matrix is called the Hessian matrix of f at p .

Note that if the third order partial derivatives of f are continuous, then the mixed second order partial derivatives of f at p are independent of the order in which they are taken, and hence $A(p)$ is a symmetric matrix. In this all the eigenvalues of $A(p)$ are real.

Theorem 4.9.1 (The Second derivative test). Let $f(t_1, t_2, \dots, t_n)$ be a real valued function in n real variables for which all third order partial order derivatives exists and are continuous. Let $p = (p_1, p_2, \dots, p_n)$ be a critical point of f , and let $A(p)$ be the Hessian matrix of f at p .

- (a) If all eigenvalues of $A(p)$ are positive, then f has a local minimum at p .
- (b) If all eigenvalues of $A(p)$ are negative, then f has a local maximum at p .
- (c) If $A(p)$ has at least one positive and at least one negative eigenvalue, then f has no local extremum at p (p is called a saddle point of f).
- (d) If $\text{rank}(A(p)) < n$ and $A(p)$ does not have both positive and negative eigenvalues, then the second derivative test is inconclusive.

Proof: If $p \neq \theta = (0, 0, \dots, 0)$, we may define a function $g: \mathbb{R}^n \rightarrow \mathbb{R}$

$$\text{by } g(t_1, t_2, \dots, t_n) = f(t_1 + p_1, t_2 + p_2, \dots, t_n + p_n) - f(p)$$

The following facts are easily verified (verify it):

1. The function f has a local maximum (minimum) at p if and only if g has a local maximum (minimum) at $\theta = (0, 0, \dots, 0)$.
2. The partial derivatives at θ are equal to the corresponding partial derivatives of f at p .
3. θ is a critical point of g .

Q. If $A(p) = [a_{ij}(p)]_{n \times n}$ then

$$4. \text{ If } A(p) = [a_{ij}(p)]_{n \times n} \text{ then } a_{ij}(p) = \frac{\partial^2 g(\theta)}{(\partial t_i)(\partial t_j)} \text{ for all } i \text{ and } j.$$

In view of these ~~for~~ facts, we may assume without loss of generality that $p = \theta$ and $f(p) = 0$.

Now we apply Taylor's theorem to f at θ to obtain the first order approximation of f around θ . We have

$$\begin{aligned} f(t_1, t_2, \dots, t_n) &= f(\theta) + \sum_{i=1}^n \frac{\partial f(\theta)}{\partial t_i} t_i + \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 f(\theta)}{(\partial t_i)(\partial t_j)} t_i t_j + S(t_1, t_2, \dots, t_n) \\ &= \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 f(\theta)}{(\partial t_i)(\partial t_j)} t_i t_j + S(t_1, t_2, \dots, t_n) \quad \dots (1) \end{aligned}$$

where S is a real valued function on \mathbb{R}^n such that

$$\lim_{x \rightarrow 0} \frac{S(x)}{\|x\|^2} = \lim_{(t_1, t_2, \dots, t_n) \rightarrow 0} \frac{S(t_1, t_2, \dots, t_n)}{t_1^2 + t_2^2 + \dots + t_n^2} = 0 \quad \dots (2)$$

Let $K: \mathbb{R}^n \rightarrow \mathbb{R}$ be the quadratic form defined by

$$K \begin{pmatrix} t_1 \\ t_2 \\ \vdots \\ t_n \end{pmatrix} = \frac{1}{2} \sum_{i,j=1}^n \frac{\partial^2 f(0)}{(\partial t_i)(\partial t_j)} t_i t_j \quad \dots (3)$$

Let H be the symmetric bilinear form corresponding to K , and β be the standard ordered basis for \mathbb{R}^n . It is easy to verify (verify it) that $\Psi_\beta(H) = \frac{1}{2} A(\beta)$. Since $A(\beta)$ is symmetric, our previous result says that

there exists an orthogonal matrix Q such that

$$Q^t A(\beta) Q = \begin{pmatrix} \lambda_1 & 0 & \dots & 0 \\ 0 & \lambda_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \lambda_n \end{pmatrix}$$

is a diagonal matrix whose diagonal entries are the eigenvalues of $A(\beta)$. Let $\gamma = \{v_1, v_2, \dots, v_n\}$ be the orthogonal basis for \mathbb{R}^n whose i th vector is the i th column of Q . Then Q is the change of coordinates matrix changing γ -coordinates into β -coordinates and by

Theorem 4.4.9

$$\Psi_\gamma(H) = Q^t \Psi_\beta(H) Q = \frac{1}{2} Q^t A(\beta) Q = \begin{pmatrix} \frac{\lambda_1}{2} & 0 & \dots & 0 \\ 0 & \frac{\lambda_2}{2} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & \frac{\lambda_n}{2} \end{pmatrix}$$

Suppose that $A(\beta)$ is not the zero matrix. Then $A(\beta)$ has non-zero eigenvalues. Choose $\epsilon > 0$ such that $\epsilon < |\lambda_i|/2$

for all $\lambda_i \neq 0$. By (2), $\exists \delta > 0$ such that for any $x \in \mathbb{R}^n$ satisfying $0 < \|x\| < \delta$, we have $|S(x)| < \varepsilon \|x\|^2$. Consider any $x \in \mathbb{R}^n$ such that $0 < \|x\| < \delta$. Then by (1) and (3)

$$|f(x) - K(x)| = |S(x)| < \varepsilon \|x\|^2$$

and hence

$$K(x) - \varepsilon \|x\|^2 < f(x) < K(x) + \varepsilon \|x\|^2 \quad \dots (4)$$

Suppose that $x = \sum_{i=1}^n s_i v_i$. Then

$$\|x\|^2 = \sum_{i=1}^n s_i^2 \quad \text{and} \quad K(x) = \frac{1}{2} \sum_{i=1}^n \lambda_i s_i^2$$

Combining these equations with (4), we obtain

$$\sum_{i=1}^n \left(\frac{1}{2} \lambda_i - \varepsilon \right) s_i^2 < f(x) < \sum_{i=1}^n \left(\frac{1}{2} \lambda_i + \varepsilon \right) s_i^2 \quad \dots (5)$$

Now suppose all eigenvalues of $A(p)$ are positive.

Then $\frac{1}{2} \lambda_i - \varepsilon > 0$ for all i and hence by the inequality of

$$(5), \quad f(0) = 0 \leq \sum_{i=1}^n \left(\frac{1}{2} \lambda_i - \varepsilon \right) s_i^2 < f(x)$$

Thus $f(0) \leq f(x)$ for $\|x\| < \delta$. So, f has a

local minimum at 0 . By a similar argument using the right inequality in (5), we have that if all ^{the} eigenvalues ~~are~~ ^{the} ~~negatives~~ of $A(p)$ are negative, then f has a local ^{maximum} ~~extension~~ at 0 . This establishes (a) and (b) of the theorem.

Next, suppose that $A(p)$ has both a positive and a negative eigenvalue, say $\lambda_i > 0$ and $\lambda_j < 0$. Then $\frac{1}{2} \lambda_i - \varepsilon > 0$