

§ Hessian of a smooth function

$$Hess(f)(X, Y) = X(Y(f)) - \nabla_X Y(f)$$

Lemma

The Hessian is a contravariant, symmetric two-tensor.

For any vector fields X and Y

1. $Hess(f)(X, Y) = Hess(f)(Y, X)$. The proof of symmetry is direct from the torsion-free assumption.

$$Hess(f)(X, Y) - Hess(f)(Y, X) = [X, Y](f) - (\nabla_X Y - \nabla_Y X)(f) = 0$$

對稱 $\nabla_X Y - \nabla_Y X = [X, Y]$ (called torsion free)

2. $Hess(f)(\phi X, \psi Y) = \phi\psi Hess(f)(X, Y)$ for all smooth functions ϕ, ψ

Other formulas for the Hessian are

1. $Hess(f)(X, Y) = \langle \nabla_X(\nabla f), Y \rangle = \nabla_X(\nabla_Y(f)) = \nabla^2 f(X, Y)$

$$\langle \nabla_X(\nabla f), Y \rangle = X(\langle \nabla f, Y \rangle) - \langle \nabla f, \nabla_X Y \rangle = X(Y(f)) - \nabla_X Y(f) = Hess(f)(X, Y)$$

2. $Hess(f)_{ij} = \partial_i \partial_j f - (\partial_k f) \Gamma_{ij}^k$ in local coordinates

$$\nabla^2 u = \begin{pmatrix} u_{x_1 x_1} & u_{x_1 x_2} & \cdots & u_{x_1 x_n} \\ u_{x_2 x_1} & u_{x_2 x_2} & \cdots & u_{x_2 x_n} \\ \vdots & \vdots & \ddots & \vdots \\ u_{x_n x_1} & u_{x_n x_2} & \cdots & u_{x_n x_n} \end{pmatrix}. \quad \text{is called the Hessian matrix of } u$$

The Laplacian Δf is defined as the trace of the Hessian :

$$\text{In local coordinates near } p, \text{ we have } \Delta f(p) = \sum_{ij} g^{ij} Hess(f)(\partial_i, \partial_j)$$

Thus, if $\{X_i\}$ is an orthonormal basis for $T_p M$ then $\Delta f(p) = \sum_i Hess(f)(X_i, X_i)$

Since $df = (\partial_r f) dx^r$ and $\nabla(dx^k) = -\Gamma_{ij}^k dx^i \otimes dx^j$, it follows that

$\nabla(df) = (\partial_i \partial_j f - (\partial_k f) \Gamma_{ij}^k) dx^i \otimes dx^j$, It is direct from the definition that

$$Hess(f)_{ij} = Hess(f)(\partial_i, \partial_j) = \partial_i \partial_j f - (\partial_k f) \Gamma_{ij}^k$$

§ 參考[[Extrema](#)]