

Higher derivatives as multilinear maps

Suppose $U \subseteq \mathbf{R}^n$ is a domain (i.e. open and connected) and $f: U \rightarrow \mathbf{R}^m$ is differentiable at all $x \in U$. As we have seen, the derivative map of f at x is a linear map of Δx , i.e. $f'(x): \mathbf{R}^n \rightarrow \mathbf{R}^m$. Making the dependence on x explicit, we get a linear-map-valued function of x , i.e. $f': U \rightarrow \text{Hom}[\mathbf{R}^n, \mathbf{R}^m]$. The derivative map of the first derivative is a linear map of say $\overline{\Delta x}$, i.e. $f''(x): \mathbf{R}^n \rightarrow \text{Hom}[\mathbf{R}^n, \mathbf{R}^m]$. We may think of this as a bilinear map of Δx and $\overline{\Delta x}$, i.e. $f''(x): \mathbf{R}^n \times \mathbf{R}^n \rightarrow \mathbf{R}^m$. Similarly we see that the k -th derivative is a k -linear map $f^{(k)}(x): (\mathbf{R}^n)^k \rightarrow \mathbf{R}^m$. It turns out that these multilinear maps are symmetric.

Cartesian representation: Writing $\Delta x = (\Delta x_1, \dots, \Delta x_n)$ we may represent the linear map $f'(x)$ by multiplication with the derivative matrix (the matrix of partial derivatives, a.k.a. the Jacobian matrix)

$$f'(x)(\Delta x) = \sum_{i=1}^n f_i(x) \Delta x_i$$

Second derivative: For simplicity assume that $n = 2$ and $m = 1$. Taking the derivative we obtain

$$\begin{aligned} f''(x)(\Delta x, \overline{\Delta x}) &= (f'(x)(\Delta x))(\overline{\Delta x}) = f'(x)(\Delta x)_1 \overline{\Delta x}_1 + f'(x)(\Delta x)_2 \overline{\Delta x}_2 \\ &= (f_1(x) \Delta x_1 + f_2(x) \Delta x_2)_1 \overline{\Delta x}_1 + (f_1(x) \Delta x_1 + f_2(x) \Delta x_2)_2 \overline{\Delta x}_2 \\ &= f_{11}(x) \Delta x_1 \overline{\Delta x}_1 + f_{21}(x) \Delta x_2 \overline{\Delta x}_1 + f_{12}(x) \Delta x_1 \overline{\Delta x}_2 + f_{22}(x) \Delta x_2 \overline{\Delta x}_2 \end{aligned}$$

The matrix representing this bilinear map is the matrix of second partial derivatives of f . Note that this is a symmetric matrix.

$$f''(x)(\Delta x, \overline{\Delta x}) = (\Delta x_1, \Delta x_2) \begin{pmatrix} f_{11}(x) & f_{12}(x) \\ f_{21}(x) & f_{22}(x) \end{pmatrix} \begin{pmatrix} \overline{\Delta x}_1 \\ \overline{\Delta x}_2 \end{pmatrix}$$

More generally with arbitrary n and m we can write

$$f''(x)(\Delta x, \overline{\Delta x}) = \sum_{j=1}^n f'(x)(\Delta x)_j \overline{\Delta x}_j = \sum_{j=1}^n \left(\sum_{i=1}^n f_i(x) \Delta x_i \right)_j \overline{\Delta x}_j = \sum_{i,j=1}^n f_{ij}(x) \Delta x_i \overline{\Delta x}_j$$

Higher derivatives: We can obtain similar formulas for higher derivatives with the symmetric tensor of higher partial derivatives.

Taylor polynomials: Suppose f is k times differentiable at x . Then $f(x + \Delta x)$ is “approximated” (as $\Delta x \rightarrow 0$) by the k -th Taylor polynomial

$$f(x + \Delta x) \approx f(x) + f'(x)(\Delta x) + \frac{1}{2} f''(x)(\Delta x, \Delta x) + \frac{1}{6} f'''(x)(\Delta x, \Delta x, \Delta x) + \dots + \frac{1}{k!} f^{(k)}(x)(\Delta x, \Delta x, \dots, \Delta x)$$

For example in cartesian coordinates with $n = 2$ and $m = 1$ we get

$$\begin{aligned} f(x_1 + \Delta x_1, x_2 + \Delta x_2) &\approx f(x) + f_1(x) \Delta x_1 + f_2(x) \Delta x_2 + \frac{1}{2} f_{11}(x) \Delta x_1^2 + f_{12}(x) \Delta x_1 \Delta x_2 + \frac{1}{2} f_{22}(x) \Delta x_2^2 \\ &+ \dots + \frac{1}{k!} \sum_{i_1, i_2, \dots, i_k=1}^n f_{i_1 i_2 \dots i_k}(x) \Delta x_{i_1} \Delta x_{i_2} \dots \Delta x_{i_k} \end{aligned}$$