## Derivative as a linear map

**Tangent space:** Let  $x \in \mathbf{R}^n$  and consider displacement vectors from x. These displacements, usually denoted  $\Delta x$ , form a vector space called the tangent space. The tangent space is just another copy of  $\mathbf{R}^n$  but with the origin shifted to x. The components of a displacement vector  $\Delta x$  with respect to the standard basis are denoted  $\Delta x_j$ .

**Slope:** Given a function  $f: \mathbf{R} \to \mathbf{R}$ , suppose we can draw a tangent line to the graph of f at a point (x, f(x)). This line does not necessarily go through the origin, but if we shift the origin to (x, f(x)) and think of the line as the graph of a function on the tangent space, it does. We get a covector — a linear map of  $\Delta x$  and denote it df. In fact,  $df(\Delta x) = m \Delta x$ , where m is the *slope* of the line.

**Osculatory approximation:** The graph of the tangent line is very close to the graph of f near x. More precisely we can say that the difference between these graphs:  $\Delta f - df(\Delta x)$ , where  $\Delta f = f(x + \Delta x) - f(x)$ , becomes very small as  $\Delta x \to 0$ , faster than  $\Delta x$  itself. Mathematically this means that as  $\Delta x \to 0$ 

$$\frac{\Delta f - df(\Delta x)}{\Delta x} = \frac{f(x + \Delta x) - f(x) - m \Delta x}{\Delta x} \to 0$$

This means that  $m = \lim_{\Delta x \to 0} \frac{f(x + \Delta x) - f(x)}{\Delta x}$ . Thus, the slope *m* is the derivative of *f* at *x* and is denoted f'(x). Taking the special case of the indentity function f(x) = x we obtain  $df(\Delta x) = \Delta x$ . In this case *df* is denoted *dx*, which is none other than the indentity map of  $\Delta x$ . We may now rewrite  $df(\Delta x) = m \Delta x = f'(x) \Delta x = f'(x) dx(\Delta x)$ , so dropping the variable  $\Delta x$ , we obtain df = f'(x) dx.

The formula df = f'(x) dx is the source of the alternate notation for the derivative  $f'(x) = \frac{df}{dx}$ .

**Linear map** df for vector variables: If  $f: \mathbb{R}^n \to \mathbb{R}^m$ , we define df to be the linear map of  $\Delta x$  such that as  $\Delta x \to 0$ .

$$\frac{\Delta f - df(\Delta x)}{|\Delta x|} \to 0$$

Note that this is a vector formula with the numerator in  $\mathbf{R}^m.$ 

**Partial derivatives, the derivative matrix:** Let us take a special case  $\Delta x = he_j$ . Then  $|\Delta x| = |he_j| = |h|$  and

$$\frac{\Delta f - df(he_j)}{|h|} \to 0, \quad \text{so} \quad \frac{f(x + he_j) - f(x) - h \, df(e_j)}{h} \to 0.$$

Therefore  $df(e_j) = \lim_{h \to 0} \frac{f(x + he_j) - f(x)}{h}$ , which the partial derivative of f with respect to  $x_j$  and is denoted  $\frac{\partial f}{\partial x_j}$ . We see that df is represented by an  $m \times n$  matrix, called the derivative matrix, whose columns are partial derivatives of f. If  $\Delta x = \sum_{j=1}^{n} h_j e_j$ , then  $dx_i(\Delta x) = dx_i\left(\sum_{j=1}^{n} h_j e_j\right) = \sum_{j=1}^{n} h_j dx_i(e_j) = h_i$ , so  $\Delta x = \sum_{j=1}^{n} dx_j(\Delta x)e_j$ . Therefore,  $df(\Delta x) = df\left(\sum_{j=1}^{n} h_j e_j\right) = \sum_{j=1}^{n} h_j df(e_j) = \sum_{j=1}^{n} df(e_j) dx_j(\Delta x) = \sum_{j=1}^{n} \frac{\partial f}{\partial x_j} dx_j(\Delta x)$ . Dropping the variable  $\Delta x$  we get  $df = \sum_{j=1}^{n} \frac{\partial f}{\partial x_j} dx_j$ .

The derivative matrix is also known as the Jacobian matrix. In the special case m = 1, the  $1 \times n$  derivative matrix may be thought of as a row vector of partial derivatives, known as the *gradient* and denoted grad f or  $\nabla f$ .

Rules of differentiation: The important rules are

- (a) Constant: d(c) = 0
- (b) Linearity: d(f+g) = df + dg
- (c) Product:  $d(f \cdot g) = df \cdot g + f \cdot dg$
- (d) Chain: d(f(g)) = f'(g) dg

Copyright 1997 Dr. Dmitry Gokhman