

## Appendix B. Calculus of Variations

We can think of a function  $y(x)$  as being an operator that, for any input value  $x$ , returns an output value  $y$ . In the same way, we can define a *functional*  $F[y]$  to be an operator that takes a function  $y(x)$  and returns an output value  $F$ . An example of a functional is the length of a curve drawn in a two-dimensional plane in which the path of the curve is defined in terms of a function. In the context of machine learning, a widely used functional is the entropy  $H[x]$  for a continuous variable  $x$  because, for any choice of probability density function  $p(x)$ , it returns a scalar value representing the entropy of  $x$  under that density. Thus, the entropy of  $p(x)$  could equally well have been written as  $H[p]$ .

A common problem in conventional calculus is to find a value of  $x$  that maximizes (or minimizes) a function  $y(x)$ . Similarly, in the calculus of variations we seek a function  $y(x)$  that maximizes (or minimizes) a functional  $F[y]$ . That is, of all possible functions  $y(x)$ , we wish to find the particular function for which the functional  $F[y]$  is a maximum (or minimum). The calculus of variations can be used, for instance, to show that the shortest path between two points is a straight line or that the maximum entropy distribution is a Gaussian.

If we were not familiar with the rules of ordinary calculus, we could evaluate a conventional derivative  $dy/dx$  by making a small change  $\epsilon$  to the variable  $x$  and then expanding in powers of  $\epsilon$ , so that

$$y(x + \epsilon) = y(x) + \frac{dy}{dx}\epsilon + \mathcal{O}(\epsilon^2) \quad (\text{B.1})$$

and finally taking the limit  $\epsilon \rightarrow 0$ . Similarly, for a function of several variables  $y(x_1, \dots, x_D)$ , the corresponding partial derivatives are defined by

$$y(x_1 + \epsilon_1, \dots, x_D + \epsilon_D) = y(x_1, \dots, x_D) + \sum_{i=1}^D \frac{\partial y}{\partial x_i} \epsilon_i + \mathcal{O}(\epsilon^2). \quad (\text{B.2})$$

The analogous definition of a functional derivative arises when we consider how much a functional  $F[y]$  changes when we make a small change  $\epsilon\eta(x)$  to the function