

## DISTANCE COVARIANCE FUNCTIONS

## Why

It is common to consider random functions whose domain is time, space, or n-dimensional space.

## Definition

Let (X, d) be a metric space. A distance covariance function  $k : X \times X \to \mathbf{R}$  is a covariance function satisfying

$$k(x,y) > k(x,y) \longleftrightarrow d(x,y) < d(x,y).$$

In other words, the covariance decreases as the distance between the arguments decreases.

## Example: squared exponential

Let  $k: X \times X \to \mathbf{R}$  be defined by

$$k(x,y) = \exp(-d(x,y)).$$

Then k is a distance covariance function. It is often called the *squared* exponential covariance function.

Let 
$$\alpha, \sigma \in \mathbf{R}$$
. Define  $k' : X \times X \to \mathbf{R}$  by

$$k'(x,y) = \alpha \exp(-d(x,y)/\sigma^2)$$

then k' is still a covariance function. In this context  $\sigma$  is often referred to as the *characteristic length-scale* of the process. The scalar  $\alpha$  is sometimes called a "prefactor" that "controls" the "overall variance."

Suppose  $(X, d) = (\mathbf{R}^n, \| \cdot \|)$ . Then the squared exponential covariance function

$$\alpha \exp(-\|x - y\|/(2\sigma^2))$$

is sometimes called the radial basis function or gaussian covariance function.<sup>1</sup> Also called an exponentiated quadratic kernel.

<sup>&</sup>lt;sup>1</sup>For reasons that will be included in future editions.

