

## Math Tools: Time Series Data

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We describe a couple ways to summarize the dynamic patterns evident in time series data: a sample of observations  $(x_t, x_2, \dots, x_T)$ . What's different about time series data is that the order matters:  $x_3$  is next to  $x_2$  and  $x_4$ , which is typically relevant to how we think about them.

We develop two tools for describing the behavior of time series variables. The first is the *autocorrelation function*, a summary of the relation between  $x_t$  and  $x_{t-k}$  for various values of  $k$ . The second is the *cross-correlation function* a summary of the relation between  $x_t$  and  $y_{t-k}$ .

### 1 Autocovariances and autocorrelations

You may recall that the sample mean is

$$\bar{x} = T^{-1} \sum_{t=1}^T x_t$$

and the variance is

$$\gamma_x(0) = T^{-1} \sum_{t=1}^T (x_t - \bar{x})^2.$$

The rational for the odd notation should be clear shortly.

Consider the covariance of  $x_t$  with  $x_{t-k}$ , for  $k$  a nonnegative integer. The sample covariance is computed

$$\gamma_x(k) = T^{-1} \sum_{t=k+1}^T (x_t - \bar{x})(x_{t-k} - \bar{x}).$$

Since we only have the observations  $x_t$  for  $t = 1, \dots, T$ , we need to start the sum at  $t = k+1$ . By longstanding convention, we nevertheless divide the sum by  $T$  rather than  $T - k$ . We could also consider negative values of  $k$ , but we'd have to adjust the range in the sum appropriately. We refer to  $\gamma_x(k)$ , a function of  $k$ , as the autocovariance function; that is, the covariances of  $x$  with itself, so to speak. When  $k = 0$ , we get the variance.

The shape of  $\gamma_x(k)$  is useful in telling us about the dynamics of  $x$ , but it's more common to scale it by  $\gamma_x(0)$  and convert it to a correlation. The autocorrelation function  $\rho_x(k)$  is defined by

$$\rho_x(k) = \gamma_x(k) / \gamma_x(0).$$

Obviously  $\rho_x(0) = 1$ :  $x_t$  is perfectly correlated with  $x_t$ . But for other values of  $k$  it can take a variety of forms.

We see, for example, that autocorrelations of equity returns are very small: returns are virtually uncorrelated over time. Interest rates, however, are very persistent: the autocorrelations decline slowly with  $k$ . You can verify other patterns in the data we used in class.

## 2 Cross-covariances and cross-correlations

We can extend the idea to the relation between two variables, say  $x$  and  $y$ . The sample *cross-covariance function* (cross meaning across two variables) is defined by

$$\gamma_{xy}(k) = T^{-1} \sum_t (x_t - \bar{x})(y_{t-k} - \bar{y}),$$

where the sum is over the appropriate range. If  $k$  is negative, we're looking at the covariance of  $x$  and future  $y$ . If  $k$  is positive, we're looking at the covariance of  $x$  and past  $y$ . Either way, we learn something about the dynamic association of  $x$  and  $y$ .

As before, it's conventional to report correlations rather than covariances. The *cross-correlation function* is

$$\rho_{xy}(k) = \frac{\gamma_{xy}(k)}{\gamma_x(0)^{1/2} \gamma_y(0)^{1/2}},$$

the covariance divided by the product of standard deviations.