

Estimating the covariance matrix

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The covariance matrix

Consider a measurement containing N samples,

$$\mathbf{x}_N = \left[\begin{array}{ccc} x_1 & \dots & x_N \end{array} \right]^T$$

The covariance matrix of \mathbf{x}_N is

$$\mathbf{R}_{\mathbf{x}} = E\left\{\mathbf{x}_{N}\mathbf{x}_{N}^{*}\right\} = \begin{bmatrix} C\{x_{1}, x_{1}\} & \dots & C\{x_{1}, x_{N}\} \\ \vdots & \ddots & \vdots \\ C\{x_{N}, x_{1}\} & \dots & C\{x_{N}, x_{N}\} \end{bmatrix}$$

$$= \begin{bmatrix} r_{x}(0) & r_{x}^{*}(1) & r_{x}^{*}(2) & \dots & r_{x}^{*}(N) \\ r_{x}(1) & r_{x}(0) & r_{x}^{*}(1) & \dots & r_{x}^{*}(N-1) \\ r_{x}(2) & r_{x}(1) & r_{x}(0) & \dots & r_{x}^{*}(N-2) \\ \vdots & & \ddots & & \vdots \\ r_{x}(N) & r_{x}(N-1) & r_{x}(N-2) & \dots & r_{x}(0) \end{bmatrix}$$

where $(\cdot)^*$ denotes the conjugate. This is a *Toeplitz* structured matrix.

This structure allows for the forming of computationally efficient algorithms. Notably, one may also express the inverse of a Toeplitz matrix in closed form!



The covariance matrix

How should one proceed to estimate $\mathbf{R}_{\mathbf{x}}$ from \mathbf{x}_{N} ?

This is not straight-forward, and there are several different ways to do so. Some of the more common include:

- $\bullet\,$ The Toeplitz-structured estimate
- The outer-product estimate

The Toeplitz-structured estimate is formed by first estimating $\hat{r}_x(k)$, typically using the *biased* estimator, and then forming $\hat{\mathbf{R}}_{\mathbf{x}}$ using the Toeplitz structure of the matrix.

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The outer-product estimate is formed by splitting \mathbf{x}_N into M subvectors of length L, such that

$$\mathbf{x}_t = \left[\begin{array}{ccc} x_t & \dots & x_{t+L-1} \end{array} \right]^T$$

where t = 1, ..., M = N - L + 1. Then, the outer-product covariance matrix estimate

$$\hat{\mathbf{R}}_{\mathbf{x}} = \frac{1}{M} \sum_{t=1}^{M} \mathbf{x}_{t} \mathbf{x}_{t}^{*}$$

Although the resulting $L \times L$ estimate is typically not a Toeplitz matrix, this is typically the preferable way to estimate $\hat{\mathbf{R}}_{\mathbf{x}}$.