

Lec4 - ACF

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Auto-Correlation Function

The auto-regressive (AR) signal model, $AR(p)$, is given by

$$\hat{y}[n] = a_1 y[n-1] + a_2 y[n-2] + \dots + a_p y[n-p] + \epsilon[n]$$

$$AR(1) \quad \hat{x}[n] = a_1 x[n-1] + \epsilon[n]$$

$$\mathcal{L}(a_1) = \sum_n (x[n] - a_1 x[n-1])^2$$

$$\frac{d\mathcal{L}(a)}{da} = 2 \sum_n (x[n] - a_1 x[n-1])(-x[n-1]) = 0$$

$$a_1 = \frac{\sum_n x[n]x[n-1]}{\sum_n x^2[n-1]}$$

Matrix formulation

$$\begin{bmatrix} x[1] \\ x[2] \\ \vdots \\ x[L-2] \\ x[L-1] \end{bmatrix} = a_1 \begin{bmatrix} x[0] \\ x[1] \\ \vdots \\ x[L-3] \\ x[L-2] \end{bmatrix}$$

$$\hat{\mathbf{y}} = a_1 \mathbf{x} \quad \mathcal{L} = \|\hat{\mathbf{y}} - a\mathbf{x}\|^2 = (\hat{\mathbf{y}} - a\mathbf{x})^T (\hat{\mathbf{y}} - a\mathbf{x})$$

$$a_1 = (\mathbf{x}^T \mathbf{x})^{-1} \mathbf{x}^T \mathbf{y} \quad \text{Least Squares (LS)}$$

Auto-correlation function

$$\hat{x}[n] = a_k x[n-k] + \epsilon[n]$$

$$a_k = \frac{\sum_n x[n]x[n-k]}{\sum_n x^2[n-k]}$$

3) Normalized auto-correlation

$$R_{xx, norm}[k] = \frac{R_{xx}[k]}{R_{xx}[0]} \lesssim a_k$$

$$R_{xx}[0] = \sum_n x^2[n] \gtrsim \sum_n x^2[n-k] = x^2[k] + \dots + x^2[L-1]$$

1) Raw auto-correlation

$$R_{xx}[k] = \sum_n x[n]x[n-k]$$

Biased auto-correlation

$$R_{xx, biased}[k] = \frac{1}{L} \sum_n x[n]x[n-k]$$

$$= \frac{1}{L} R_{xx}[k]$$

4) Unbiased auto-correlation

$$\frac{1}{L} \sum_n x^2[n] \approx \frac{1}{L-k} \sum_n x^2[n-k]$$

$$\frac{x^2[0] + x^2[1] + \dots + x^2[L-1]}{L} \approx \frac{x^2[k] + \dots + x^2[L-1]}{L-k}$$

$$\frac{L}{L-k} R_{xx}[k] = \frac{L}{L-k} R_{xx, norm}[k] \approx a_k$$

$$R_{xx, biased}[k] = \frac{1}{L-k} \sum_n x[n]x[n-k]$$

Correlation Coefficient Interpretation

$$\mathcal{L}_{min}(a_k) = \sum_{n=0}^{L-1} x^2[n] - a_k \sum_{n=0}^{L-1} x[n]x[n-k]$$

$$= R_{xx}[0] - a_k R_{xx}[k]$$

$$\sqrt{R_{xx}[0] \left(1 - \left(\frac{R_{xx}[k]}{R_{xx}[0]} \right)^2 \right)}$$

Properties The signal energy is given by

$$E_x = \sum_n x^2[n] = R_{xx}[0]$$

and it is also the higher value of ACF,

$$R_{xx}[0] \geq R_{xx}[k].$$

The corresponding average power is given by

$$R_{xx}[0] \left(1 - \left(\frac{R_{xx}[k]}{R_{xx}[0]} \right)^2 \right) \rightarrow a_k \cdot R_{xx}[0]$$

$$= R_{xx}[0] \left(1 - R_{xx,norm}^2[k] \right) \leftarrow \rho_{xx}[k]$$

The value of $\rho_{xx}[k]$ is termed *correlation coefficient* between $x[n]$ and $x[n-k]$,

$$\rho_{xx}[k] \approx \frac{L}{L-k} R_{xx,norm}[k] \approx R_{xx,norm}[k]. \quad (20.23)$$

$$|\rho_{xx}[k]| \leq 1 \quad 0 \leq \mathcal{L}_{min}(a_k) \leq \sum_{n=0}^{L-1} x^2[n] = R_{xx}[0]$$

MSE and RMSE For example, the corresponding MSE and RMSE metrics are given by

$$MSE(a_k) = \frac{1}{L} \mathcal{L}_{min}(a_k) \quad (20.27a)$$

$$RMSE(a_k) = \sqrt{\frac{1}{L} \mathcal{L}_{min}(a_k)} \quad (20.27b)$$

Correlation time One of the way to quantify the time ahead of the linear predictability, the correlation time is used. The correlation time is defined by the smallest time, k_c , that satisfies,

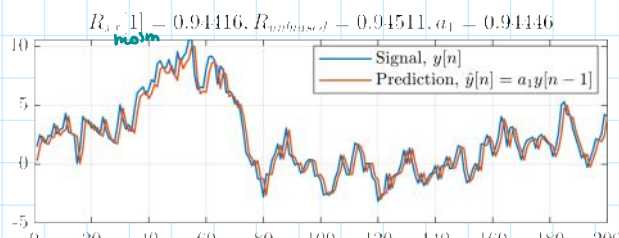
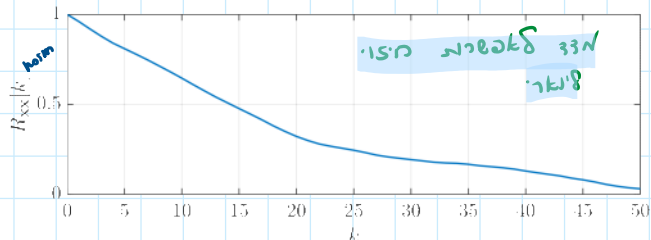
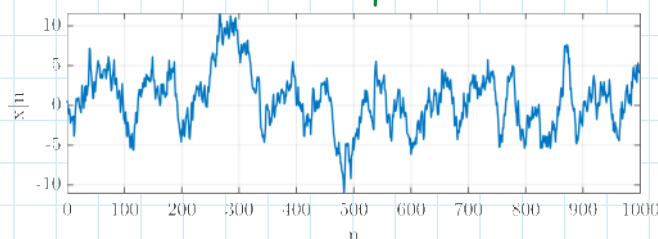
$$\rho_{xx}[k_c] = 0.5 \text{ or } 0.1 \text{ or } \exp(-1) \approx 0.36 \quad (20.28)$$

The decision threshold depends on the field of application. Practically, $\rho_{xx}[k > k_c] \approx 0$ is assumed.

למשל: $\rho_{xx}[k_c] = 0.5$ או 0.1 או $\exp(-1) \approx 0.36$

Matlab

הצגת גרף של $R_{xx}[k]$



$R_{xx}[0] \geq R_{xx}[k]$.
The corresponding **average power** is given by

$$P_x = \frac{1}{L} \sum_n x^2[n] = R_{xx,biasd}[0]$$

The ACF has inherent time symmetry,

$$R_{xx}[k] = R_{xx}[-k]$$

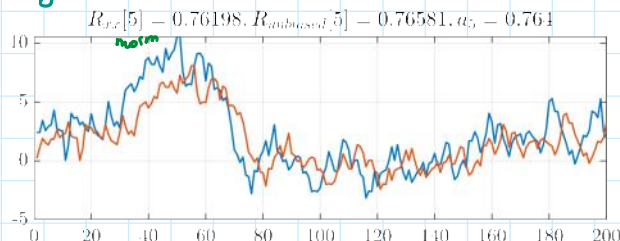
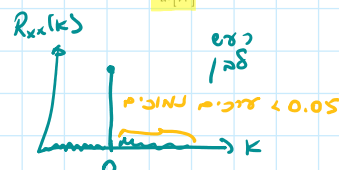
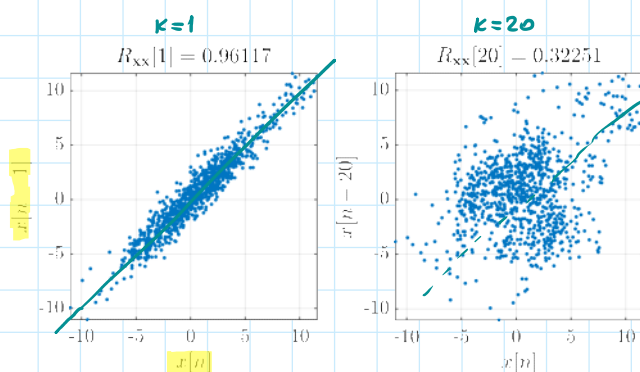
$$k = -(L-1), \dots, 0, 1, \dots, L-1$$

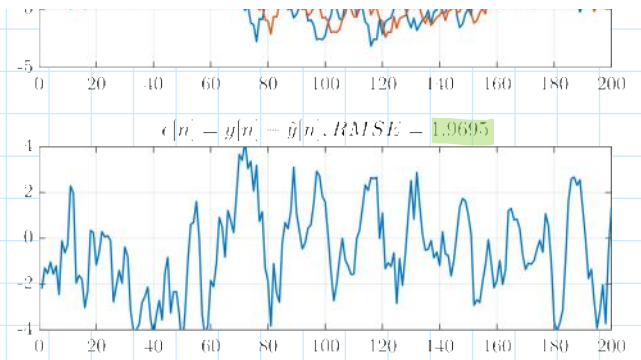
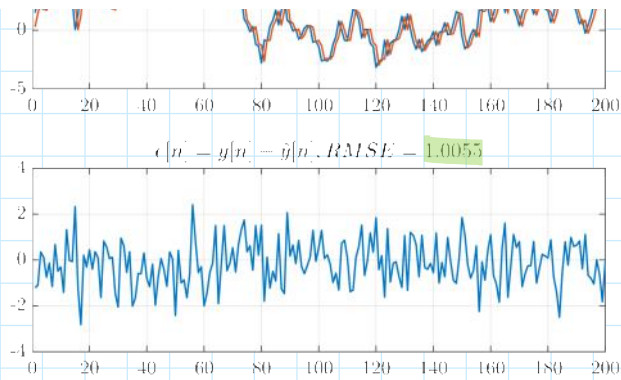
20.1.3 Auto-covariance

For simplicity, a zero-average, $\bar{x}[n] = 0$, was assumed. When the signals is non-zero mean, the subtraction of signal average from the signal, $x[n] = x[n] - \bar{x}[n]$ before auto-correlation calculation is termed as auto-covariance.

$$\bar{x}[n] = \frac{1}{L} \sum_{n=0}^{L-1} x[n] \neq 0$$

`[R,lags] = xcorr(y,50,'coeff');`





חיסול האות 'לנצוק'

$$\begin{aligned}\mathcal{L}_{min}(a_k) &= \sum_{n=0}^{L-1} x^2[n] - a_k \sum_{n=0}^{L-1} x[n]x[n-k] \\ &= R_{xx}[0] - a_k R_{xx}[k] \\ &\lesssim R_{xx}[0] \left(1 - \left(\frac{R_{xx}[k]}{R_{xx}[0]} \right)^2 \right)\end{aligned}$$

$$3) \text{ חיסול המקור} = R_{xx}[0] \left(1 - R_{xx,norm}^2[k] \right)$$

$$1) \text{ חיסול מספר ישר} \mathcal{L}_{min}(a_k) = \mathbf{e}^T \mathbf{e}$$

חיסול בית