

Lecture #2: LLSE — geometric interpretation

Gustaf Hendeby
gustaf.hendeby@liu.se

Version: 2023-11-11

1 Geometric Interpretation

Recall from Lecture #1, the llse. of x , \hat{x} is given by

$$\hat{x} = \Sigma_{xy} \Sigma_{yy}^{-1} y = K_o y$$

Interpretation:

$$K_o = \Sigma_{xy} \Sigma_{yy}^{-1} \Leftrightarrow K_o E[yy^*] = E[xy^*] \Leftrightarrow E[(x - K_o y)y^*] = 0$$

Thus, if the random variables are viewed as vectors, with the inner product defined as $\langle x, y \rangle = E[xy^*]$, then $E[(x - K_o y)y^*]$ has the geometric interpretation

$$\langle x - K_o y, y \rangle = 0 \Leftrightarrow x - K_o y \perp y,$$

that is, the estimation error is orthogonal to observations!

Two questions related to the geometric interpretation:

1. Which vector space?

Hilbert space, i.e., a complete normed linear vector space.

2. Is $\langle x, y \rangle = E[xy^*]$ a proper inner product?

- Linearity: $\langle \alpha x + \beta y, z \rangle = \alpha \langle x, z \rangle + \beta \langle y, z \rangle$

$$E[(\alpha x + \beta y)z^*] = \alpha E[xz^*] + \beta E[yz^*] \quad \text{Check!}$$

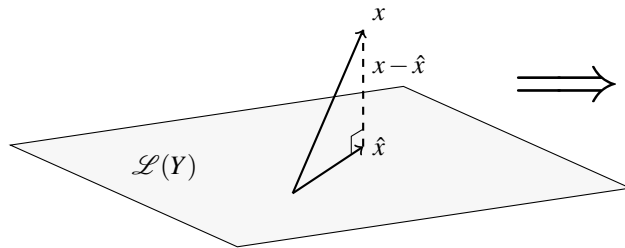
- Symmetry: $\langle x, y \rangle = \langle y, x \rangle^*$

$$E[xy^*] = E[(yx^*)^*] = E[yx^*]^* \quad \text{Check!}$$

- Non-degeneracy: $\langle x, x \rangle \geq 0$, i.e., positive semi-definite, and $\langle x, x \rangle = 0$ if and only if $x = 0$.
 $E[xx^*] = \Sigma_{xx}$: Covariance matrices are per definition always positive (semi-)definite.

1.1 Geometric Derivation of LLSE

Find a vector \hat{x} in the linear space spanned by $\{y_1, \dots, y_N\}$ such that $\|x - \hat{x}\|$ is minimized.



$$Y = \begin{pmatrix} y_1 \\ \vdots \\ y_N \end{pmatrix}, \quad \hat{x} = K_o y$$

$$\begin{aligned} x - \hat{x} &\perp Y \\ \langle x - \hat{x}, y \rangle &= 0 \\ E[xy^*] &= K_o E[yy^*] \\ K_o &= \Sigma_{xy} \Sigma_{yy}^{-1} \end{aligned}$$

2 Winer filtering (WF)

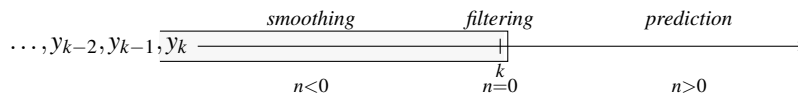
Given the observations $\{y_i\}_{i=-\infty}^k$, find the llse. of x_{k+n} .

Assumptions: x_k and y_k are scalar processes, that are jointly stationary with exponentially bounded (cross-)covariance, i.e., $|r_{xx}(k)| < K\alpha^{|k|} \forall k, K > 0, 0 < \alpha < 1$, then a spectrum exists.

That is, find

$$\hat{x}_{k+n} = \sum_{i=0}^{\infty} h_{k,i} y_{k-i},$$

where $h_{k,i}$ is possibly time varying filter coefficients subjects to $E[(x_{k+n} - \hat{x}_{k+n})^2]$ is minimized.



Orthogonality properties gives, for all $j \leq k$:

$$\begin{aligned} x_{k+n} - \hat{x}_{k+n} &\perp y_i, \\ \langle x_{k+n} - \sum_{i=0}^{\infty} h_{k,i} y_{k-i}, y_j \rangle &= 0, \\ E[x_{k+n} y_j^*] &= \sum_{i=0}^{\infty} h_{k,i} E[y_{k-i} y_j^*], \\ r_{xy}(k+n-j) &= \sum_{j=0}^{\infty} h_{k,i} r_{yy}(k-i-j). \end{aligned}$$

Make a change variable, $k - j \rightarrow \ell$, gives

$$r_{xy}(\ell+n) = \sum_{i=0}^{\infty} h_{\ell+j,i} r_{yy}(\ell-i), \quad \forall \ell \geq 0.$$

Note, neither $r_{xy}(\ell+n)$ nor $r_{yy}(\ell-i)$ depend on j ; hence, $h_{\ell+j,i} = h_i$, i.e., the filter coefficients are time invariant.

Wiener-Hopf equation:

$$r_{xy}(\ell+n) = \sum_{i=a}^b h_i r_{yy}(\ell-i), \quad \forall \ell \geq 0.$$

- If:
- $\sum_{-\infty}^{\infty}$: Non-causal *Wiener filter* (WF, z -transform), easy to solve.
 - \sum_a^b : FIR (*finite impulse response*) WF (linear system equation), easy to solve.
 - \sum_0^{∞} : Casual WF.
 - $\sum_{-\infty}^{-1}$: Anti-casual WF.

2.1 Spectral Factorization

The spectrum of a signal y_k

$$\Phi_{yy}(\omega) = \tilde{\Phi}_{yy}(z) \Big|_{z=e^{i\omega}}$$

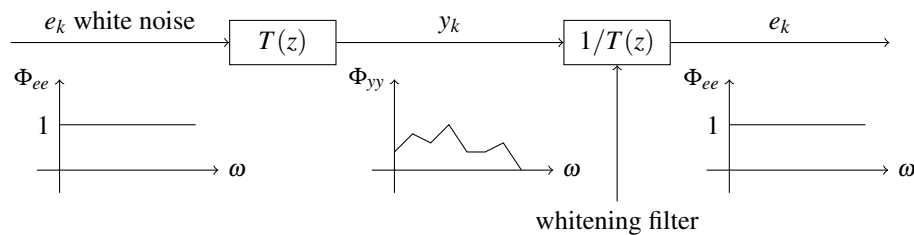
$$\tilde{\Phi}_{yy}(z) = \underbrace{\mathcal{L}\{r_{yy}(k)\}}_{\text{acf}} = \sum_{k=-\infty}^{\infty} r_{yy}(k) z^{-k}$$

Note, $r_{yy}(k) = r_{yy}(-k) \Rightarrow \tilde{\Phi}_{yy}(z) = \tilde{\Phi}_{yy}(z^{-1})$ which implies that $\tilde{\Phi}_{yy}(z)$ has symmetry with respect to mirroring in the unit circle. Hence, if $z = r_i$ has a pole (zero) in the unit circle, then $z = r_i^{-1}$ is also a pole (zero).

If $\tilde{\Phi}_{yy}(z)$ has no poles or zeroes on the unit circle, i.e., $0 < \Phi_{yy}(\omega) < \infty \forall \omega$, then

$$\tilde{\Phi}_{yy}(z) = \underbrace{\sigma_e \frac{\prod_{i=1}^m (z - r_i)}{\prod_{j=1}^b (z - p_j)}}_{=T(z), \text{ stable, causal}} \cdot \underbrace{\sigma_e \frac{\prod_{i=1}^m (z^{-1} - r_i^*)}{\prod_{j=1}^b (z^{-1} - p_j^*)}}_{=T^*(z^{-*}), \text{ stable, anti-causal}},$$

assuming $|r_i| < 1$, $|p_i| < 1$, and $\sigma_e^2 > 0$.



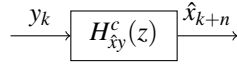
2.2 Additive Decomposition

Let the sequence $\{f_k\}$ have a \mathcal{L} -transform that exist in an annulus containing the unit circle. Then

$$F(z) = \sum_{k=-\infty}^{\infty} f_k z^{-k} = \underbrace{\sum_{k=0}^{\infty} f_k z^{-k}}_{[F(z)]_+ \text{ casual part}} + \underbrace{\sum_{k=-\infty}^{-1} f_k z^{-k}}_{[F(z)]_- \text{ strictly anti-causal part}}$$

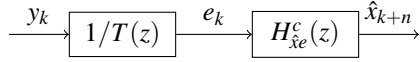
2.3 Solving the Wiener-Hopf equation

Original problem:



$$r_{xy}(\ell+n) = \sum_{i=0}^{\infty} h_i r_{yy}(\ell-i), \quad \forall \ell \geq 0$$

New problem:



Different filter coefficients

$$r_{xe}(\ell+n) = \sum_{i=0}^{\infty} \bar{h}_i r_{ee}(\ell-i) = \sum_{i=0}^{\infty} \bar{h}_i \delta(\ell-i), \quad \forall \ell \geq 0$$

$$\implies \bar{h}_i = \begin{cases} r_{xe}(i+n), & i \geq 0 \\ 0, & i < 0 \end{cases}$$

$$\implies H_{xe}^c(z) = [\tilde{\Phi}_{xe}(z)z^n]_+$$

Putting it all together

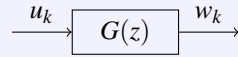
$$w_k = \begin{pmatrix} e_k \\ x_k \end{pmatrix}$$

$$\longrightarrow \tilde{\Phi}_{ww} = \begin{pmatrix} \tilde{\Phi}_{ee} & \tilde{\Phi}_{ex} \\ \tilde{\Phi}_{xe} & \tilde{\Phi}_{xx} \end{pmatrix}$$

$$u_k = \begin{pmatrix} y_k \\ x_k \end{pmatrix}$$

$$\longrightarrow \tilde{\Phi}_{uu} = \begin{pmatrix} \tilde{\Phi}_{yy} & \tilde{\Phi}_{yx} \\ \tilde{\Phi}_{xy} & \tilde{\Phi}_{xx} \end{pmatrix}$$

Super formula:



$$\tilde{\Phi}_{ww} = G(z)\tilde{\Phi}_{ww}G^*(z^{-*})$$

$$G(z) = \begin{pmatrix} 1/T(z) & 0 \\ 0 & 1 \end{pmatrix}$$

$$\implies \tilde{\Phi}_{xe} = \tilde{\Phi}_{xy}/T^*(z^{-*})$$

$$\implies H_{xe}^c(z) = \left[\frac{z^n \tilde{\Phi}_{xy}(z)}{T^*(z^{-*})} \right]_+$$

$$\implies H_{xy}^c(z) = \frac{1}{T(z)} \cdot \left[\frac{z^n \tilde{\Phi}_{xy}(z)}{T^*(z^{-*})} \right]_+$$

Note: A factor z^{-n} is in some books added to $H_{xy}^c(z)$. Without this factor (as given above) $\hat{x}_{k+n|k} = H_{xy}^c(\Delta)y_k$ and with the factor $\hat{x}_{k+n|k} = \Delta^{-n}H_{xy}^c(\Delta)y_{k+n}$. Be sure to know which convention you are adhering to, and both works just fine.