

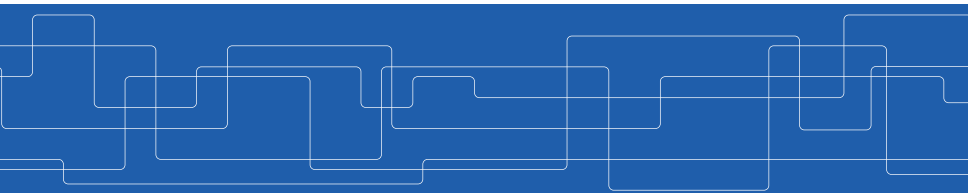


Linear Estimation in Krein Spaces

Braghadeesh Lakshminarayanan

Division of Decision and Control Systems,
KTH Royal Institute of Technology,
Stockholm, Sweden

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Motivation: H^∞ Problem

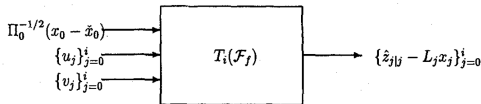
$$\begin{cases} x_{i+1} = F_i x_i + G_i u_i, & x_0 \\ y_i = H_i x_i + v_i, & i \geq 0 \end{cases}$$

- ▶ Goal: To estimate some arbitrary linear combination of the states, say

$$z_i = L_i x_i$$

- ▶ A Posteriori estimate: $\hat{z}_{i|i} = \mathcal{F}_f(y_0, y_1, \dots, y_i)$
- ▶ How do we gauge the "quality" of the above estimate?

Motivation: H^∞ Problem



H^∞ Problem

Find estimation strategies $\check{z}_{i|j} = \mathcal{F}_f(y_0, y_1, \dots, y_i)$ that achieve $\|T_i(\mathcal{F}_f)\|_\infty < \gamma_f$ ($\gamma_f > 0$)

$$\|T_i(\mathcal{F}_f)\|_\infty = \sup_{x_0, u \in h_2, v \in h_2} \frac{\sum_{j=0}^i e_{f,j}^* e_{f,j}}{(x_0 - \check{x}_0)^* \Pi_0^{-1} (x_0 - \check{x}_0) + \sum_{j=0}^i u_j^* u_j + \sum_{j=0}^i v_j^* v_j}$$

[2] "Linear estimation in Krein spaces. II. Applications" Hassibi, B.; Sayed, A.H.; Kailath, T. Automatic Control, IEEE Transactions on Volume: 41 1, Jan. 1996, Page(s): 34 -49

Motivation: H^∞ Problem

$$\|T\|_\infty := \sup_{u \in h_2, u \neq 0} \frac{\|Tu\|_2}{\|u\|_2},$$

where $\|u\|_2$ is the h_2 -norm of the causal sequence $\{u_k\}$, i.e., $\|u\|_2^2 = \sum_{k=0}^{\infty} u_k^* u_k$

- ▶ Interpretation: Maximum energy gain from the input u to the output y
- ▶ Idea: Filter \mathcal{F}_f to H^∞ can be obtained as certain Kalman filter via suitable construction of state-space model
- ▶ Caution: Projections cannot be realized in Hilbert space unlike standard Kalman filter.

Remedy?

Projection in Krein Space!

- ▶ Projections in Krein Space often lead to indefinite quadratic forms \implies need additional conditions to ensure uniqueness and minimum.



Outline

Krein Space

Projections and Quadratic Forms

Back to H^∞ : Krein Space Solution

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Krein Space

An abstract vector space $\{\mathcal{K}, \langle \cdot, \cdot \rangle\}$ that satisfies the following requirements is called a Krein Space:

i) \mathcal{K} is a linear space over \mathcal{C} , the complex numbers.

ii) $\langle \mathbf{y}, \mathbf{x} \rangle = \langle \mathbf{x}, \mathbf{y} \rangle^*$,

iii) $\langle a\mathbf{x} + b\mathbf{y}, \mathbf{z} \rangle = a\langle \mathbf{x}, \mathbf{z} \rangle + b\langle \mathbf{y}, \mathbf{z} \rangle$

for any $\mathbf{x}, \mathbf{y}, \mathbf{z} \in \mathcal{K}$, $a, b \in \mathcal{C}$

iv) The vector space \mathcal{K} admits a direct orthogonal sum decomposition

$$\mathcal{K} = \mathcal{K}_+ \oplus \mathcal{K}_-$$

such that $\{\mathcal{K}_+, \langle \cdot, \cdot \rangle\}$ and $\{\mathcal{K}_-, -\langle \cdot, \cdot \rangle\}$ are Hilbert spaces, and

$$\langle \mathbf{x}, \mathbf{y} \rangle = 0$$

for any $\mathbf{x} \in \mathcal{K}_+$ and $\mathbf{y} \in \mathcal{K}_-$.

Note: If \mathbf{x}, \mathbf{y} are stochastic variables, then $\langle \mathbf{x}, \mathbf{y} \rangle := \mathbb{E}[\mathbf{xy}^*]$

[1] "Linear estimation in Krein spaces. I. Theory" Hassibi, B.; Sayed, A.H.; Kailath, T. Automatic Control, IEEE Transactions on Volume: 41 1, Jan. 1996, Page(s): 18 -33



Krein Space

Krein space looks like Hilbert space except that

- ▶ $\langle \mathbf{x}, \mathbf{x} \rangle = 0 \not\Rightarrow \mathbf{x} = \mathbf{0}$ (\mathbf{x} is called Neutral vector)
- ▶ There exists $\mathbf{x} \neq \mathbf{0}$ in a linear subspace (M) of \mathcal{K} such that $\mathbf{x} \perp M$ (\mathbf{x} is called Isotropic vector)
- ▶ $\langle \mathbf{x}, \mathbf{x} \rangle$ can be negative



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Definition

Given \mathbf{z} , $\{\mathbf{y}_0, \mathbf{y}_1, \dots, \mathbf{y}_N\}$ in \mathcal{K} , $\hat{\mathbf{z}}$ is projection of \mathbf{z} onto $\mathcal{L}\{\mathbf{y}_0, \mathbf{y}_1, \dots, \mathbf{y}_N\}$ if

$$\mathbf{z} = \hat{\mathbf{z}} + \tilde{\mathbf{z}}$$

where $\hat{\mathbf{z}} \in \mathcal{L}\{\mathbf{y}_0, \dots, \mathbf{y}_N\}$ and $\tilde{\mathbf{z}}$ satisfies the orthogonality condition

$$\tilde{\mathbf{z}} \perp \mathcal{L}\{\mathbf{y}_0, \dots, \mathbf{y}_N\}$$

or equivalently, $\langle \tilde{\mathbf{z}}, \mathbf{y}_i \rangle = 0$ for $i = 0, 1, \dots, N$.

Remark:

- ▶ Projections always exist and unique in Hilbert space
- ▶ In Krein space, existence and uniqueness of projection require additional conditions



Projections in Krein Spaces

Lemma

If the Gramian matrix $R_y = \langle \mathbf{y}, \mathbf{y} \rangle$ is nonsingular, then the projection of \mathbf{z} onto $\mathcal{L}\{\mathbf{y}\}$ exists, is unique, and is given by

$$\hat{\mathbf{z}} = \langle \mathbf{z}, \mathbf{y} \rangle \langle \mathbf{y}, \mathbf{y} \rangle^{-1} \mathbf{y} = R_{zy} R_y^{-1} \mathbf{y}.$$

So, standing assumption: R_y is nonsingular



Projections and Quadratic Forms

- ▶ In Hilbert space, projections minimize certain quadratic forms
- ▶ In Krein space, projections stationarize certain quadratic forms
- ▶ Need additional condition for stationary point to be minimum

To this end, we look at two closely related problems,

- ▶ Stochastic minimization problem
- ▶ A partially equivalent deterministic problem

Stochastic Minimization Problem

A natural quadratic form to study: error Gramian

$$P(k) = \langle \mathbf{z} - \mathbf{k}^* \mathbf{y}, \mathbf{z} - \mathbf{k}^* \mathbf{y} \rangle,$$

- i) $\{\mathbf{y}_0, \dots, \mathbf{y}_N\}$: collection of elements in a Krein space \mathcal{K} with indefinite inner product $\langle \cdot, \cdot \rangle$,
- ii) $\mathbf{z} = \text{col} \{z_0, \dots, z_M\}$: some column vector of elements in \mathcal{K} ,
- iii) $\mathbf{k}^* \mathbf{y}$ an arbitrary linear combination of $\{\mathbf{y}_0, \dots, \mathbf{y}_N\}$. $\mathbf{y} = \text{col} \{\mathbf{y}_0, \dots, \mathbf{y}_N\}$

$$P(k) = \|\mathbf{z} - \hat{\mathbf{z}}\|_{\mathcal{K}}^2 + \|\hat{\mathbf{z}} - \mathbf{k}^* \mathbf{y}\|_{\mathcal{K}}^2$$

where $\hat{\mathbf{z}} = \mathbf{k}_0^* \mathbf{y} = R_{zy} R_y^{-1}$ is the projection of \mathbf{z} onto $\mathcal{L}(\mathbf{y})$

Stochastic Minimization Problem

- ▶ $\|\hat{\mathbf{z}} - k^* \mathbf{y}\|^2 = \|k_0^* \mathbf{y} - k^* \mathbf{y}\|^2 = 0$, even if $k_0 \neq k$.
- ▶ $\|k_0^* \mathbf{y} - k^* \mathbf{y}\|^2$ could be negative, $P(k)$ may not be minimized by choosing $k = k_0$

Theorem 1

When R_y is nonsingular, k_0 , the unique coefficient matrix in the projection of \mathbf{z} onto $\mathcal{L}\{\mathbf{y}\}$, $\hat{\mathbf{z}} = k_0^* \mathbf{y}$, $k_0 = R_y^{-1} R_{yz}$ yields the unique stationary point of the error Gramian

$$\begin{aligned} P(k) &\triangleq \langle \mathbf{z} - k^* \mathbf{y}, \mathbf{z} - k^* \mathbf{y} \rangle \\ &= \begin{bmatrix} I & -k^* \end{bmatrix} \begin{bmatrix} R_z & R_{zy} \\ R_{yz} & R_y \end{bmatrix} \begin{bmatrix} I \\ -k \end{bmatrix} \end{aligned}$$

Further, k_0 is a unique minimum iff $R_y > 0$ i.e., R_y is not only nonsingular but also positive definite.

A Partially Equivalent Deterministic Problem

- Scalar second-order form:

$$J(z, y) \triangleq \begin{bmatrix} z^* & y^* \end{bmatrix} \begin{bmatrix} R_z & R_{zy} \\ R_{yz} & R_y \end{bmatrix}^{-1} \begin{bmatrix} z \\ y \end{bmatrix}$$

Theorem 2

Suppose both R_y and the block matrix in above equation are nonsingular. Then

a) The stationary point z_0 of $J(z, y)$ over z is given by

$$z_0 = R_{zy}R_y^{-1}y.$$

b) The value of $J(z, y)$ at the stationary point is

$$J(z_0, y) = y^*R_y^{-1}y$$

Further, z_0 is a minimum iff $R_z - R_{zy}R_y^{-1}R_{yz} > 0$.



Implications of Theorem 1 and 2

- ▶ Stationary point for the scalar second-order form is “same” as that in Theorem 1 for Krein space projection of vector \mathbf{z} onto $\mathcal{L}(\mathbf{y})$
- ▶ H^∞ problems will lead directly to certain indefinite quadratic forms:
 - ▶ Theorem 1 provides algorithm to find stationary point via corresponding Krein-space projection
 - ▶ Theorem 2 is used to check if stationary point obtained above is indeed the minimum



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Back to H^∞ : Krein Space Solution

- ▶ Recall H^∞ problem: $\|T_i(\mathcal{F}_f)\|_\infty < \gamma_f$

- ▶ For all nonzero $\left\{x_0, \{u_j\}_{j=0}^i, \{v_j\}_{j=0}^i\right\}$

$$\frac{\sum_{j=0}^i |\check{z}_{j|j} - L_j x_j|^2}{(x_0 - \check{x}_0)^* \Pi_0^{-1} (x_0 - \check{x}_0) + \sum_{j=0}^i |u_j|^2 + \sum_{j=0}^i |y_j - H_j x_j|^2} < \gamma_f^2$$

\implies for all $k \leq i$

$$\frac{\sum_{j=0}^k |\check{z}_{j|j} - L_j x_j|^2}{(x_0 - \check{x}_0)^* \Pi_0^{-1} (x_0 - \check{x}_0) + \sum_{j=0}^k |u_j|^2 + \sum_{j=0}^k |y_k - H_j x_j|^2} < \gamma_f^2.$$

- ▶ The above inequality is an indefinite quadratic form

Back to H^∞ : Krein Space Solution

Lemma 3

$\|T_i(\mathcal{F}_f)\|_\infty < \gamma_f$ iff there exists $\check{z}_{k|k} = \mathcal{F}_f(y_0, \dots, y_k)$ (for all $0 \leq k \leq i$) such that for all complex vectors x_0 , for all causal sequences $\{u_j\}_{j=0}^i$, and for all nonzero causal sequences $\{y_j\}_{j=0}^i$, the scalar quadratic form

$$\begin{aligned}
 J_{f,k}(x_0, u_0, \dots, u_k, y_0, \dots, y_k) &= x_0^* \Pi_0^{-1} x_0 + \sum_{j=0}^k u_j^* u_j \\
 &\quad + \sum_{j=0}^k (y_j - H_j x_j)^* (y_j - H_j x_j) \\
 &\quad - \gamma_f^{-2} \sum_{j=0}^k (\check{z}_{j|j} - L_j x_j)^* (\check{z}_{j|j} - L_j x_j)
 \end{aligned}$$

satisfies $J_{f,k}(x_0, u_0, \dots, u_k, y_0, \dots, y_k) > 0$ for all $0 \leq k \leq i$

Back to H^∞ : Krein Space Solution

To show that $J_{f,k}(x_0, u_0, \dots, u_k, y_0, \dots, y_k) > 0$ for all $0 \leq k \leq i$, following Lemma is used

Lemma 4

The scalar quadratic forms $J_{f,k}(x_0, u_0, \dots, u_k, y_0, \dots, y_k)$ satisfy the conditions above iff, for all $0 \leq k \leq i$

i) $J_{f,k}(x_0, u_0, \dots, u_k, y_0, \dots, y_k)$ has a minimum with respect to $\{x_0, u_0, u_1, \dots, u_k\}$.

ii) The $\{\check{z}_{k|k}\}_{k=0}^i$ can be chosen such that the value of $J_{f,k}(x_0, u_0, \dots, u_k, y_0, \dots, y_k)$ at this minimum is positive, viz.

$$\min_{\{x_0, u_0, \dots, u_k\}} J_{f,k}(x_0, u_0, \dots, u_k, y_0, \dots, y_k) > 0$$



Implications of Lemma 3 and 4

- ▶ Lemma 3 provides necessary and sufficient condition for solution to H^∞ problem.
- ▶ Lemma 4 guarantees that stationary point of quadratic form is a minimum
- ▶ Lemma 4 also guarantees positivity of the quadratic form

To find the stationary point, we make use of Krein-space projection.

- ▶ Need a corresponding state-space model

Krein Space State Space Model

The following Krein-space system is introduced:

$$\begin{cases} \mathbf{x}_{j+1} = F_j \mathbf{x}_j + G_j \mathbf{u}_j \\ \begin{bmatrix} \mathbf{y}_j \\ \check{z}_{j|i} \end{bmatrix} = \begin{bmatrix} H_j \\ L_j \end{bmatrix} \mathbf{x}_j + \mathbf{v}_j \end{cases}$$

with

$$\left\langle \begin{bmatrix} \mathbf{x}_0 \\ \mathbf{u}_j \\ \mathbf{v}_j \end{bmatrix}, \begin{bmatrix} \mathbf{x}_0 \\ \mathbf{u}_k \\ \mathbf{v}_k \end{bmatrix} \right\rangle = \begin{bmatrix} \Pi_0 & 0 & 0 \\ 0 & I \delta_{jk} & 0 \\ 0 & 0 & \begin{bmatrix} I & 0 \\ 0 & -\gamma_f^2 I \end{bmatrix} \delta_{jk} \end{bmatrix}$$

- ▶ Corresponding Kalman filter recursion gives a posteriori estimate $\check{z}_{i|i} = \mathcal{F}_f(y_0, y_1, \dots, y_i)$ that satisfies $\|T_i(\mathcal{F}_f)\|_\infty < \gamma_f$



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Conclusion

- ▶ H^∞ problem leads to indefinite quadratic form
- ▶ Indefinite quadratic forms are handled well by Krein-space projection
- ▶ Recursive solution by making use of Kalman filter theory on Krein-space state-space model