

The geometry of low-rank Kalman filters

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Introduction: proof of concept

- **The natural metric of the cone of positive definite matrices** allows to analyse the convergence properties of the Kalman filter through the contraction properties of Riccati flow.
- We introduced in 2009¹ a new metric on the set of positive semi-definite matrices of fixed rank that generalizes the natural metric of the cone.
- *This metric is appropriate to analyse the **contraction properties of a low-rank** version of the **Riccati flow**.*

¹Riemannian metric and geometric mean for positive semidefinite matrices of fixed rank. Bonnabel and Sepulchre. SIAM J. Matrix Anal Appl. 2009.

Outline

- 1 Contraction properties of the Riccati flow
- 2 Low rank Kalman filtering
- 3 An invariant metric on the set of fixed rank psd matrices
- 4 Contraction properties of the low-rank Riccati flow

Kalman filter, discrete Riccati equation

System

$$\begin{aligned}x_t &= A_t x_{t-1} + B_t w_t \\y_t &= C_t x_t + v_t, \quad t = 0, 1, \dots\end{aligned}$$

Kalman filter

$$\begin{aligned}\hat{x}_t &= (A_t - P_t C_t^T C_t) \hat{x}_{t-1} + P_t C_t^T y_t \\P_t &= \Phi_t(P_{t-1})\end{aligned}$$

Discrete Riccati Equation

$$\Phi_t(P) = (A_t P A_t^T + B_t B_t^T)(I + C_t^T C_t B_t B_t^T + C_t^T C_t A_t P A_t^T)^{-1}$$

Mapping Φ_t is **positive** on the cone $P_+(n)$.

Natural metric of $P_+(n)$

Group action of $GL(n)$ on $P_+(n)$ via congruence:

$$\varphi_A(P) = A^T P A$$

Reductive homogeneous space: $P_+(n)$ admits a $GL(n)$ -invariant metric.

Metric based on Euclidian metric at Identity + invariance

$$g_P(D_1, D_2) = \text{Tr}(P^{-1/2} D_1 P^{-1} D_2 P^{-1/2})$$

Geodesic distance: $d(P, Q) = (\sum_k \log^2(\lambda_k))^{1/2}$ where $\det(PQ^{-1} - \lambda_k I) = 0$.

Invariance to congruence and inversion (=isometries)

$$d(P, Q) = d(A^T P A, A^T Q A) = d(P^{-1}, Q^{-1})$$

Contraction of the Ricatti equation

Proposition [Bougerol (1993)] for $B \succ 0$ and $C \succ 0$, the map

$$\begin{aligned}\Phi(P) &= (APA^T + B)(C(APA^T + B) + I)^{-1} \\ &= (C + (APA^T + B)^{-1})^{-1}\end{aligned}$$

is a contraction on the Riemmanian manifold $(P_+(n), g)$

$$\exists \gamma < 1 \quad \forall P_1, P_2 \in P_+(n) : d(\Phi(P_1), \Phi(P_2)) \leq \gamma d(P_1, P_2)$$

Proof idea: $\Phi(\cdot)$ decomposes as

$$(\cdot)^{-1} \circ (\cdot + C) \circ (\cdot)^{-1} \circ (A \cdot A^T) \circ (\cdot + A^{-1}BA^{-T})$$

Composition of additions and **isometries** !

Left to prove: $P \mapsto P + C$ with $C \succ 0$ is a contraction (logical).

In continuous time

Kalman-Bucy filter

$$\frac{d}{dt}\hat{x} = (A - PC'(HH')^{-1}C)\hat{x} + PC'(HH')^{-1}y \quad (1)$$

$$\frac{d}{dt}P = \Phi_t(P) = AP + PA' + GG' - PC'(HH')^{-1}CP \quad (2)$$

Contraction property in the sense of Lohmiller and Slotine for the natural metric $\|\delta P\|_g^2 = g_P(\delta P, \delta P) = \text{Tr}(P^{-1}\delta P P^{-1}\delta P)$

$$\frac{d}{dt}\delta P = A\delta P + \delta P A' - \delta P C'(HH')^{-1}CP - PC'(HH')^{-1}C\delta P$$

$$\frac{d}{dt}\|\delta P\|_g^2 \leq -2\text{Tr}((P^{-1}\delta P P^{-1}\delta P)GG'P^{-1}) \leq -2\lambda\|\delta P\|_g^2$$

with $\lambda = \mu/p_{\max}$ where μ is a lower bound on GG' , and p_{\max} is an upper bound on P .

Contraction of Kalman filter

- **Desired property of a filter:** exponential forgetting of the initial condition
- Rooted in invariance properties of the natural metric
- The same property holds for the continuous-time Riccati flow

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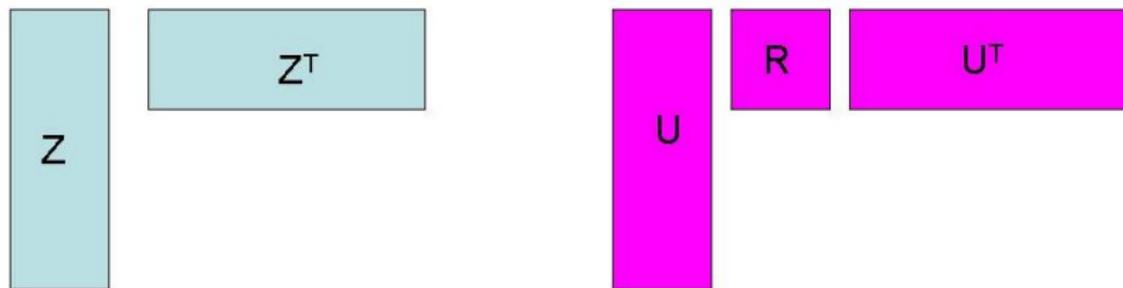
Low rank Kalman filter

- Propagating the covariance matrix $P(t)$ is expensive in large-scale applications
- Choose $P(0)$ low-rank and propagate (a low-rank approximation of) $P(t)$ instead
- Quite popular in **oceanography** (e.g. SEEK Filter, Pham et al. 1997)
- Meets applications in **EKF SLAM** (see e.g. the SEIF)
- Natural metric is **not** defined on the facets. Geometry? Contraction?

Factorizations of P of rank $r \ll n$

Square-root factorization $P = ZZ^T$ with $Z \in \mathbb{R}_*^{n \times r}$: most popular, in particular in early numerical implementations of KF.

Polar factorization $P = URU^T$ with $(U, R) \in \text{St}(r, n) \times \text{P}_+(r)$: meaningful as U defines a subspace and R an ellipsoid.



Low rank Riccati equation

Riccati equation in polar factorization

$$\frac{d}{dt} URU' = AURU' + URU'A' + GG' - URU'C'(HH')^{-1}CURU'$$

which is **not** rank-preserving.

Modified rank-preserving Riccati equation: $GG' = \mu^2 UU'$

$$\frac{d}{dt} URU' = AURU' + URU'A' + \mu^2 UU' - URU'C'(HH')^{-1}CURU'$$

which is an alternative to the usual assumption $G = 0$.

Low rank Riccati equation

The update decomposes into the triangular system

$$\frac{d}{dt}U = (I - UU')AU \quad (3)$$

$$\frac{d}{dt}R = A_U R + R A_U + \mu^2 I - R C'_U (H H')^{-1} C_U R \quad (4)$$

with $A_U = U'AU$, $C_U = CU$

Subspace update (3) known as Oja flow which tracks the dominant subspace of A .

Low-rank cone update (4) is the projected Riccati equation.

Suggests to use a **specific metric** to study the contraction properties ?!

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An invariant metric

Quotient manifold of psd matrices of fixed rank

$$S^+(r, n) \cong (\text{St}(r, n) \times P_+(r))/O(r)$$

as $P = URU^T = (UO)(O^T R O)(O^T U^T)$ for any $O \in O(r)$

Tangent space: $(\delta U, \delta R)$ tangent to $\text{St}(r, n) \times P_+(r)$ where $\delta U = U_\perp B$, i.e. is \perp to the equivalence classes $\{UO\}$.

Proposed metric: sum of infinitesimal distances between subspaces and low-rank factors of $P_+(r)$

$$g((\delta U_1, \delta R_1), (\delta U_2, \delta R_2)) = \text{Tr}(\delta U_1^T \delta U_2) + \text{Tr}(R^{-1} \delta R_1 R^{-1} \delta R_2)$$

Invariant to rotations, scalings, pseudo-inversion.

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Contraction properties

Oja flow **contracting** around the r -dominant subspace of $A^s = \frac{A+A'}{2}$ in the sense of Stiefel metric. Letting $\delta U = U_{\perp} B$

$$\frac{d}{dt} \text{Tr}(\delta U' \delta U) = 2 \text{Tr}(B B' A_{\perp}^s - B' B A_r^s) \leq -\lambda_U \text{Tr}(B' B) = -\lambda_U \text{Tr}(\delta U' \delta U)$$

Low-rank cone flow **contracting** in the sense of natural metric

$$\frac{d}{dt} \text{Tr}(R^{-1} \delta R R^{-1} \delta R) \leq -\lambda_R \frac{d}{dt} \text{Tr}(R^{-1} \delta R R^{-1} \delta R)$$

Low rank Riccati equation **contracting** in the sense of the proposed metric around the dominant subspace of A^s

$$\begin{aligned} \frac{d}{dt} g((\delta U, \delta R), (\delta U, \delta R)) &= \frac{d}{dt} [\text{Tr}(\delta U' \delta U) + \text{Tr}(R^{-1} \delta R R^{-1} \delta R)] \\ &\leq -\lambda_U \text{Tr}(\delta U' \delta U) - \lambda_R \text{Tr}(R^{-1} \delta R R^{-1} \delta R) \\ &\leq -\lambda g((\delta U, \delta R), (\delta U, \delta R)) \end{aligned}$$

Contraction and convergence

$$\frac{d}{dt}U = (I - UU')AU \quad (5)$$

$$\frac{d}{dt}R = A_U R + R A_U + \mu^2 I - R C'_U (H H')^{-1} C_U R \quad (6)$$

For A symmetric $U \rightarrow U_r$. Thus $(A_U, C_U) \rightarrow (A_r, C_r)$, no peaking. Thus

*The low-rank Riccati equation is **eventually contracting** in the sense of our metric with uniform rate.*

For $A - A^T \neq 0$, U does not converge (can rotate forever). For topological reasons it can **not** contract everywhere.

Conclusion

- The Hilbert and natural metrics are useful to analyse contraction and convergence properties of positive maps
- Unfortunately they are not defined on the facets of $P_+(n)$
- One way to study the geometry of the facets is to fix the rank via $P = URU^T$. Our Riemannian geometry

$$S_+(r, n) \approx (St(r, n) \times P_+(r))/O(r)$$

is insightful in several problems involving low-rank approximations of PSD matrices

- Here: it retains some of the contraction properties of the Riccati flow.
- Elsewhere: on $S_+(r, n)$ matrix means (filtering), online identification, optimization.