Dynamic radioastronomic image reconstruction using Kalman filter under ionospheric perturbations

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Part I

Introduction

Part II

Notations and problem

1 Notations

1.1 Hadamard product

Hadamard product, denoted \odot , is the element-wise product of same sized matrices :

$$\forall (\mathsf{A},\mathsf{B}) \in \left(\mathbb{C}^{m \times n}\right)^2, \quad \mathsf{A} \odot \mathsf{B} \in \mathbb{C}^{m \times n} \quad \text{and} \quad \forall (i,j) \in [\![1;m]\!] \times [\![1;n]\!], \, (\mathsf{A} \odot \mathsf{B})_{i,j} = \mathsf{A}_{i,j} \times \mathsf{B}_{i,j}$$

We can further introduce Hadamard inversion:

$$\forall \mathbf{A} \in \left(\mathbb{C} \setminus \{0\}\right)^{m \times n}, \quad \mathbf{A}^{\circ - 1} \in \left(\mathbb{C} \setminus \{0\}\right)^{m \times n} \quad \text{and} \quad \forall (i, j) \in [1; m] \times [1; n], \left(\mathbf{A}^{\circ - 1}\right)_{i, j} = \left(\mathbf{A}_{i, j}\right)^{-1}$$

And notation ∅ being Hadamard division:

$$\forall (\mathbf{A}, \mathbf{B}) \in \mathbb{C}^{m \times n} \times (\mathbb{C} \setminus \mathbf{0})^{m \times n}, \quad \mathbf{A} \otimes \mathbf{B} \in \mathbb{C}^{m \times n} \quad \text{and} \quad \forall (i, j) \in [1; m] \times [1; n], (\mathbf{A} \otimes \mathbf{B})_{i, j} = \frac{\mathbf{A}_{i, j}}{\mathbf{B}_{i, j}}$$

1.2 Norms and distances

In order to compute distances between matrices, we note $|| \bullet ||_F$ Frobenius norm :

$$\mathbf{A} \in \mathcal{M}_{m,n}(\mathbb{K}) \quad ||\mathbf{A}||_{\mathrm{F}} := \sqrt{\mathrm{tr}\big\{\mathbf{A}\mathbf{A}^{\mathrm{H}}\big\}} = \sqrt{\sum_{\substack{1 \leq i \leq m \\ 1 \leq j \leq n}} \left|\mathbf{A}_{ij}\right|^2}$$

Inducing the distance d_1 :

$$\forall (A, B) \in (\mathcal{M}_{m,n}(\mathbb{K}))^2$$
 $d_1(A, B) = \frac{||A - B||_F}{mn}$

In order to measure influence of a specific element, the ??? distance d_2 is the following:

$$\forall (\mathbf{A}, \mathbf{B}) \in \left(\mathcal{M}_{m,n}(\mathbb{K})\right)^2 \quad d_2(\mathbf{A}, \mathbf{B}) = \sum_{\substack{1 \le i \le m \\ 1 \le j \le n}} \left| \mathbf{A}_{ij} - \mathbf{B}_{i,j} \right|$$

2 Evolution model and Kalman filtering

We consider a set $\{x_k\}$ of n vector as theoretical images taken at a regular time interval. Those images are measured by an antenna mesh which UV plane will be written H. Measurements vectors will be designated by n measures $\{y_k\}$.

We will consider the following model to describe time related evolutions and measure equation:

$$\begin{cases} x_k = Ax_{k-1} + w_{k-1} \\ y_k = Hx_k + v_k \end{cases}$$

Where:

- A is the linear evolutionary model between time k-1 and k, supposing evolution is time invariant
- $\{w_k\}$ are the additive evolutionary errors
- $\{v_k\}$ are the measurements noises

In order to build images dynamically with better precision, a Kalman filter has been implemented.

This filter requires assumptions about

Part III

Reliability of Kalman Filter under perturbations

- 1 Perturbations sources
- 2 Effects
- 2.1 Position
- 2.2 Direction

Part IV

Experimental correction

- 1 implementation
- 2 Measures