

RFI mitigation options for Uniboard²

R. Weber (Station de radioastronomie de Nançay / Université d'Orléans)

C.Dumez-Viou (Station de radioastronomie de Nançay)

S. Changuel (Station de radioastronomie de Nançay)



- **The options**
- **Spatial processing :**
 - the context
 - The cyclostationarity concept
 - First example : Cyclic detectors
 - Second example : Estimation and subtraction
 - Third example : cyclic spatial filtering
- **Conclusions**

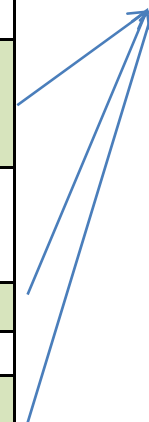
RFI mitigation options for Uniboard²



UNIVERSITÉ D'ORLÉANS

Application		EC man month (mm) Deliverable : matlab golden model
Starting time		3 mm
Digital receiver		
	Impulse detector	3 mm
	Cyclostationary detector	3 mm
	Kurtosis detector	3 mm
	FFT2D + radon transform	4 mm
Beamformer		
	Pre -beamformer Cyclostationary detector	4mm
	Post-beamformer cyclostationary detector	3mm
	Spatial filtering	???
Correlator		
	Cyclostationary detector	6 mm
Pulsar machine		
	Upgrade of Uniboard design	3 mm

Spatial processing





- **The options**
- **Spatial processing :**
 - **the context**
 - **The cyclostationarity concept**
 - **First example : Cyclic detectors**
 - **Second example : Estimation and subtraction**
 - **Third example : cyclic spatial filtering**
- **Conclusions**

Classic approach



ASTRON

Hypothesis : $\mathbf{R}_z \approx \underbrace{\mathbf{A}_r \mathbf{R}_r \mathbf{A}_r^H}_{\mathbf{R}_{\text{rfi}}} + \underbrace{\sigma_n^2 \mathbf{I}}_{\mathbf{R}_{\text{noise}}}$

$\mathbf{A}_r = \begin{matrix} \text{M antennas} \\ \left[\begin{matrix} \uparrow \\ \mathbf{a}_k \\ \downarrow \end{matrix} \right] \\ \text{K}_1 \text{ RFI} \end{matrix}$

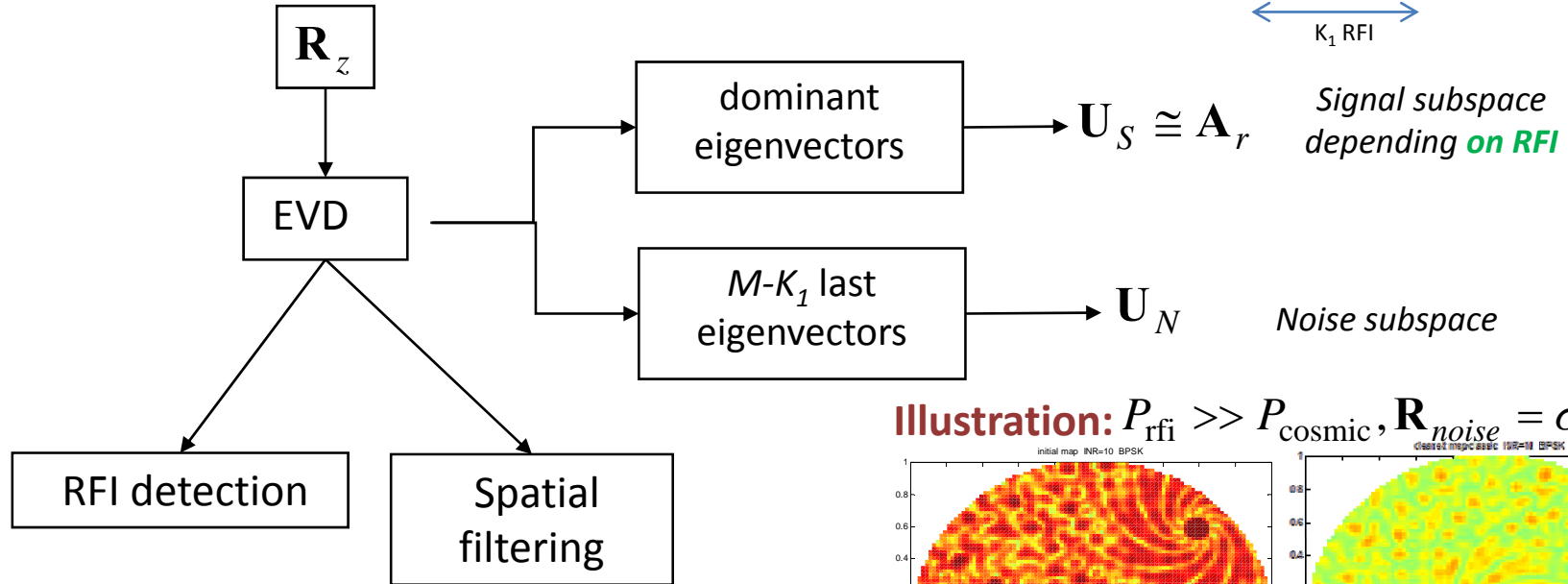
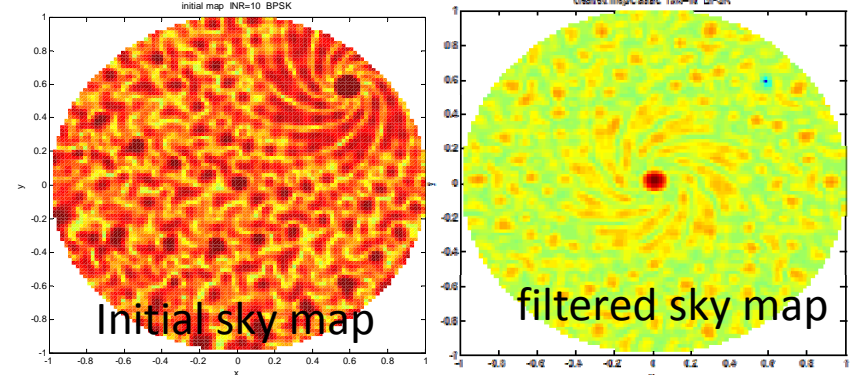


Illustration: $P_{\text{rfi}} \gg P_{\text{cosmic}}, \mathbf{R}_{\text{noise}} = \sigma_n^2 \mathbf{I}$



(see papers from A. Leshem, A.J. Boonstra, A.J van der Veen)

General case issue



ASTRON

$$\mathbf{R}_z = \underbrace{\mathbf{A}_r \mathbf{R}_r \mathbf{A}_r^H}_{\mathbf{R}_{\text{rfi}}} + \underbrace{\mathbf{A}_s \mathbf{R}_s \mathbf{A}_s^H}_{\mathbf{R}_{\text{cosmic}}} + \underbrace{\mathbf{R}_n}_{\mathbf{R}_{\text{noise}}} \times \mathbf{R}_{\text{RFI}} + \sigma_n^2 \mathbf{I}$$

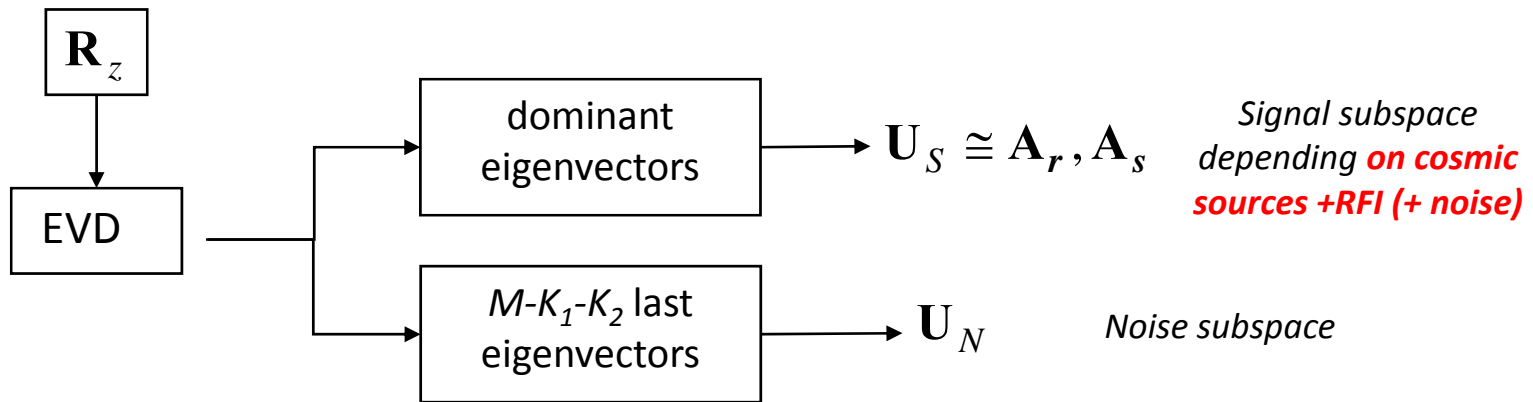
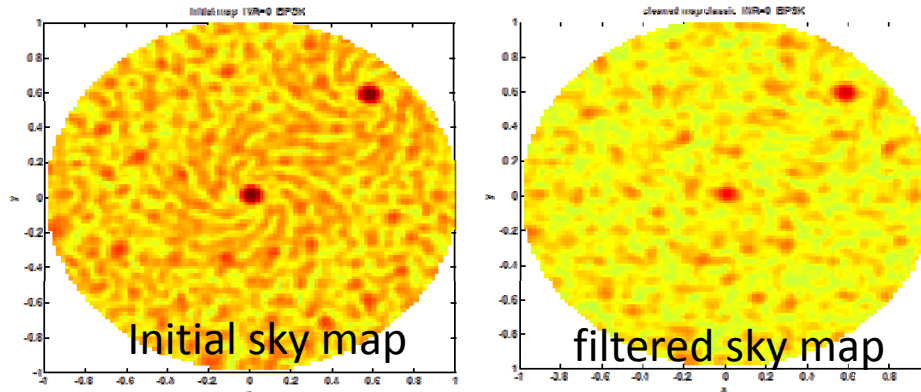


Illustration with :

$$P_{\text{rfi}} \approx P_{\text{cosmic}}$$

$$\mathbf{R}_{\text{noise}} = \sigma_n^2 \mathbf{I}$$



Uniboard meeting, 12-13 October 2010, Bordeaux

Outline



- The options
- **Spatial processing :**
 - the context
 - **The cyclostationarity concept**
 - First example : Cyclic detectors
 - Second example : Estimation and subtraction
 - Third example : cyclic spatial filtering
- Conclusions

Cyclostationarity concept



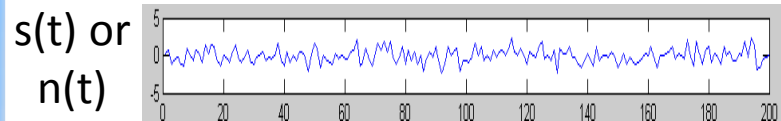
UNIVERSITÉ D'ORLÉANS

ASTRON

Stationary process

≈ statistics time-independent

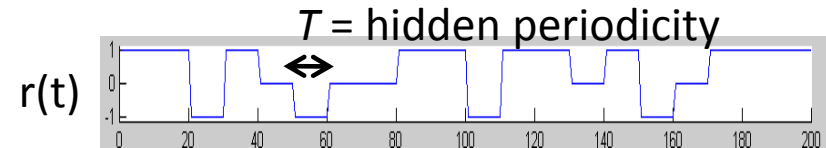
Example : second order statistics



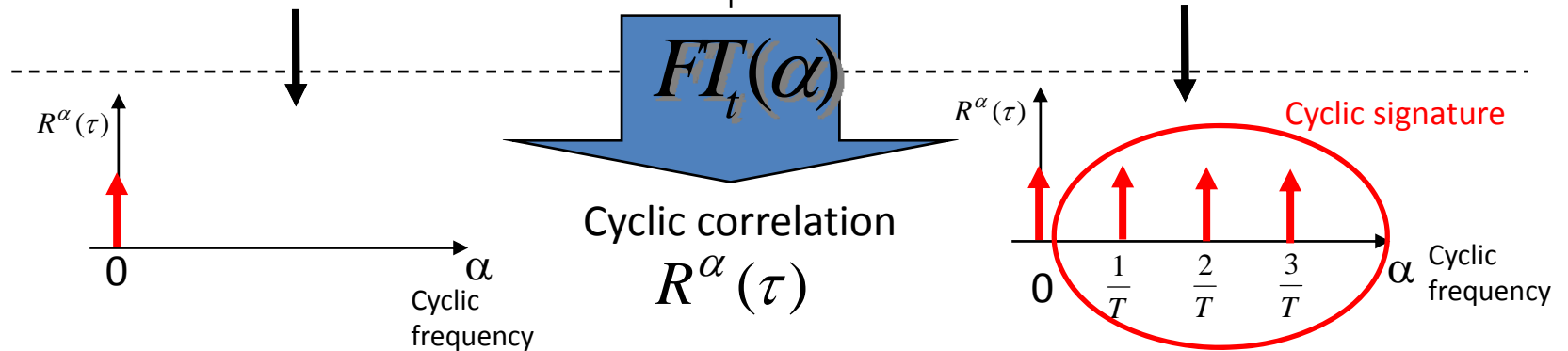
$$R(t, \tau) = R(\tau)$$

Cyclostationary process

≈ statistics are periodic



$$R(t + T, \tau) = R(t, \tau)$$



In practice :
$$R_z^\alpha(\tau) = \left\langle z(t + \tau/2) z^*(t - \tau/2) \exp(-j2\pi\alpha t) \right\rangle_\infty$$

➔ Gardner and Giannakis

The multidimensional case



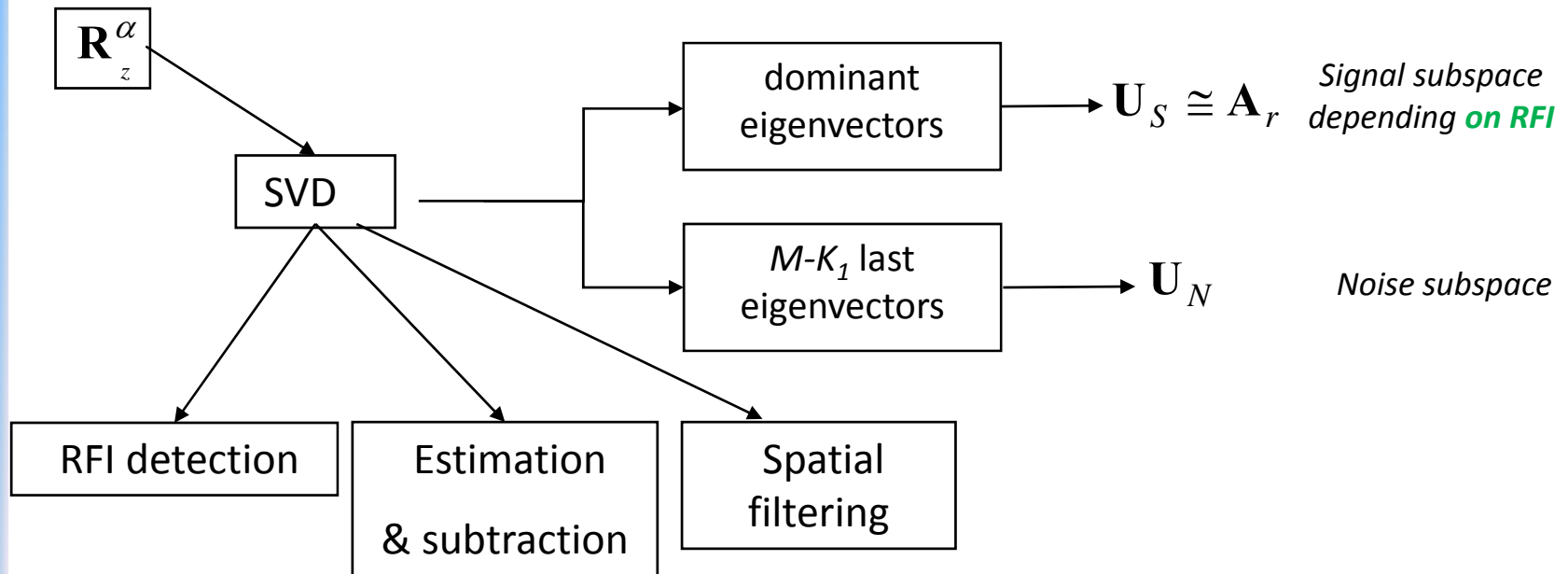
UNIVERSITE D'ORLEANS

ASTRON

$$\mathbf{R}_z^\alpha = \left\langle \mathbf{z} \left(t + \frac{\tau}{2} \right) \mathbf{z}^H \left(t - \frac{\tau}{2} \right) \exp(-j2\pi\alpha t) \right\rangle_\infty \quad \text{Cyclic correlation matrix}$$

$$= \underbrace{\mathbf{A}_r \mathbf{R}_r^\alpha(\tau) \mathbf{A}_r^H}_{\mathbf{R}_{RFI}^\alpha(\tau)} + \mathbf{A}_s \mathbf{R}_s^\alpha(\tau) \mathbf{A}_s^H + \mathbf{R}_n^\alpha(\tau)$$

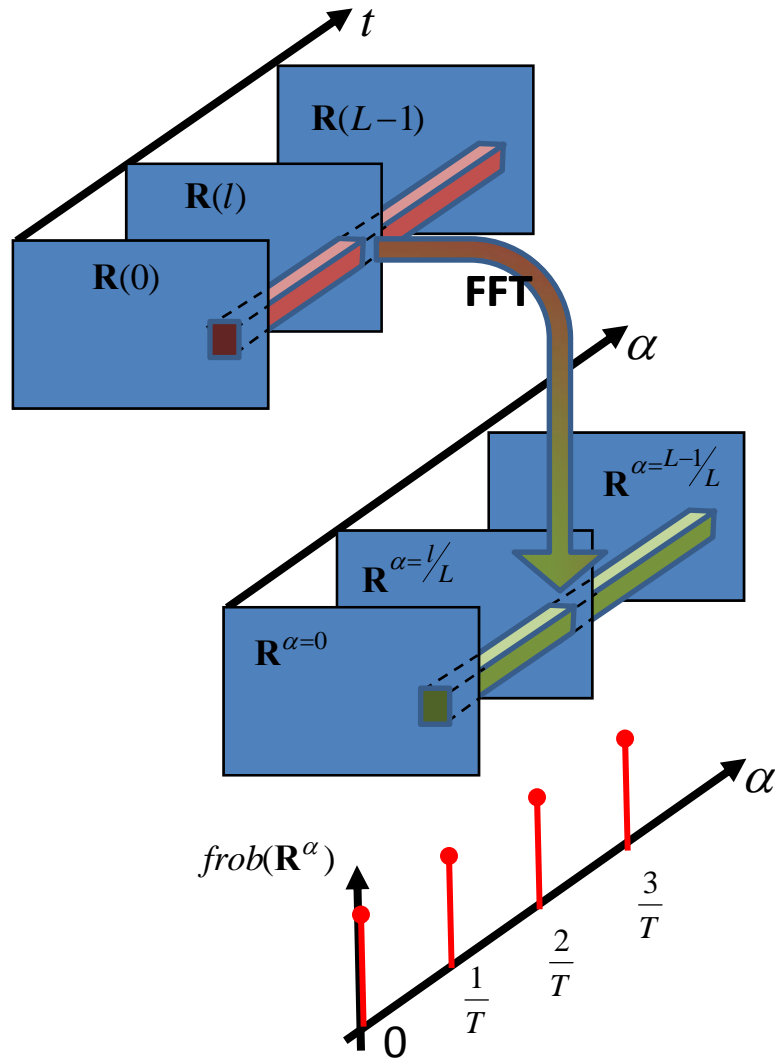
0
0





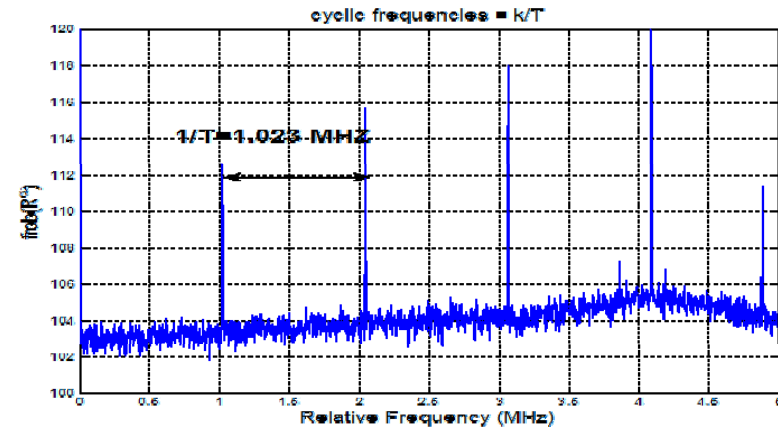
- The options
- **Spatial processing :**
 - the context
 - The cyclostationarity concept
 - **First example : Cyclic detectors**
 - **Second example : Estimation and subtraction**
 - **Third example : cyclic spatial filtering**
- **Conclusions**

Cyclic detectors



	AM	BPSK	M-PSK	M-QAM
α for R^α	none	$1/T_{symbol}$	$1/T_{symbol}$	$1/T_{symbol}$
α for \bar{R}^α	$2f_0$	$2f_0 + k/T_{sym}$	none	none

Example at Westerbork telescope with GPS data



Several implementations:

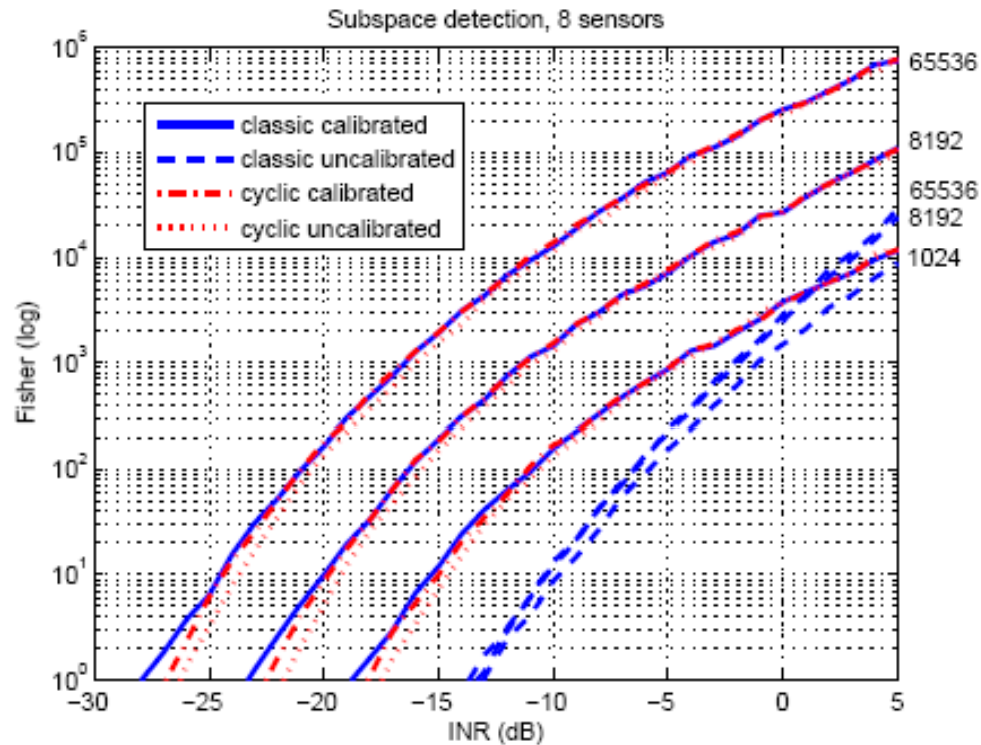
- Blind detector
- $frob(R^{\alpha=k/T}) = \sum \lambda_k^2$
- $\max(\lambda_k)$

Cyclic detector performance



ASTRON

AM RFI, no cosmic sources
Power fluctuations between antennas = 20%





- The options
- **Spatial processing :**
 - the context
 - The cyclostationarity concept
 - First example : Cyclic detectors
 - **Second example : Estimation and subtraction**
 - Third example : cyclic spatial filtering
- Conclusions

Estimation & subtraction (1)



Consider 1 BPSK RFI : $r(t) = \sum_k c_k h(t - kT - t_0) e^{j2\pi f_0 t + j\varphi_0}$

1) $\overline{\mathbf{R}}_z^{\alpha = 2f_0} = \mathbf{a}_r \mathbf{a}_r^T \left(\frac{\sigma_r^2}{T} R_h \right) e^{j2\pi\varphi_0}$

2) SVD

3)

$\mathbf{R}_z = \underbrace{\mathbf{a}_r \mathbf{a}_r^H \left(\frac{\sigma_r^2}{T} R_h \right)}_{4) \hat{\mathbf{R}}_{\text{rfi}}} + \mathbf{R}_{\text{cosmic}} + \mathbf{R}_{\text{noise}}$

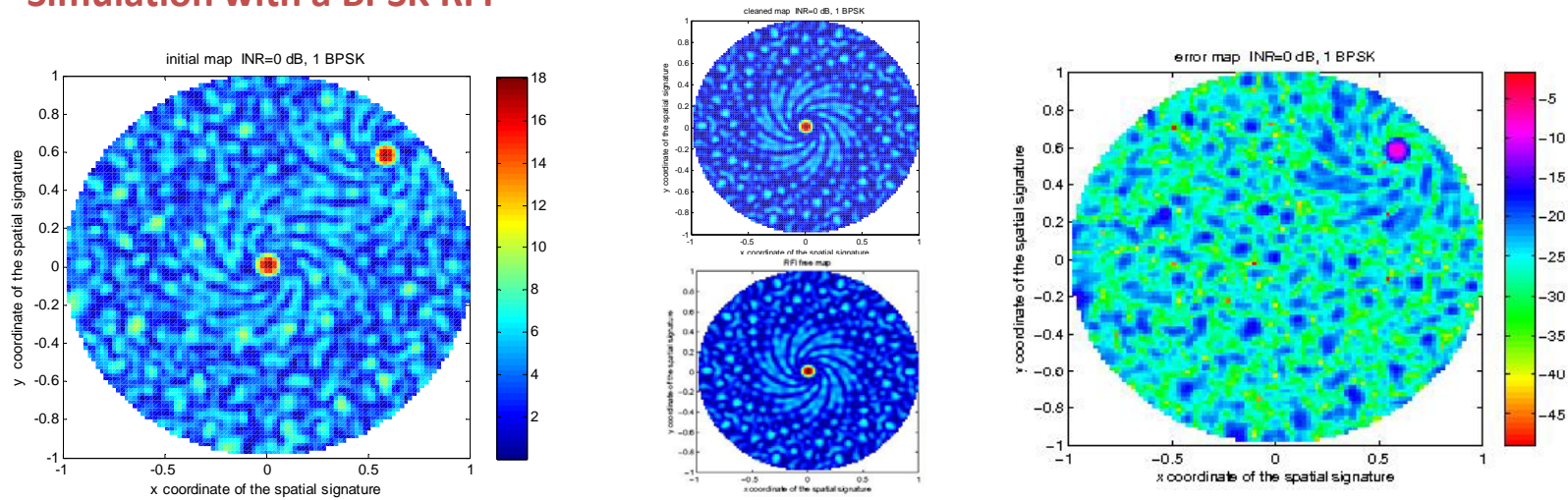
5) $\mathbf{R}_z^{\text{clean}} = \mathbf{R}_z - \hat{\mathbf{R}}_{\text{rfi}}$



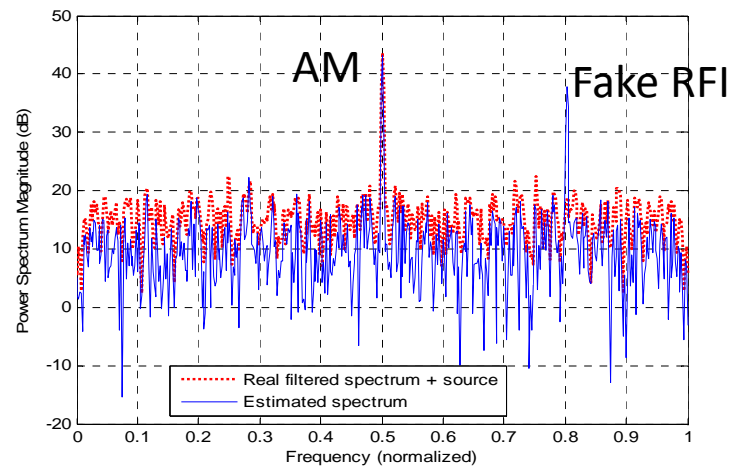
OK for AM, BPSK + QAM if $h(t)$ known

Estimation & subtraction (2)

Simulation with a BPSK RFI



Example of an AM RFI in the LOFAR band





- The options
- **Spatial processing :**
 - the context
 - The cyclostationarity concept
 - First example : Cyclic detectors
 - Second example : Estimation and subtraction
 - **Third example : cyclic spatial filtering**
- Conclusions

Cyclic Spatial Filtering (1)



ASTRON

Spatial filtering:

- 1) projector $\mathbf{P} = \mathbf{I} - \mathbf{A}_r (\mathbf{A}_r^H \mathbf{A}_r)^{-1} \mathbf{A}_r^H$
- 2) cleaning: $\mathbf{R}_{clean} = \mathbf{P} \mathbf{R} \mathbf{P}$

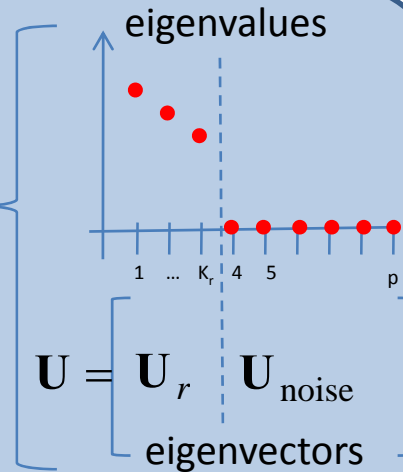
Estimation of \mathbf{A}_r :

1) SVD : $\mathbf{R}^\alpha = \mathbf{U} \mathbf{\Lambda} \mathbf{V}^H$

2) Extract the RFI subspace : \mathbf{U}_r

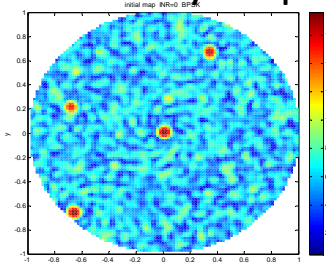
3) Since $\mathbf{U}_r = span \{ \mathbf{A}_r \}$

➔ Projector $\mathbf{P} = \mathbf{I} - \mathbf{U}_r (\mathbf{U}_r^H \mathbf{U}_r)^{-1} \mathbf{U}_r^H$

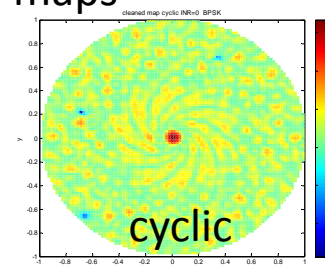
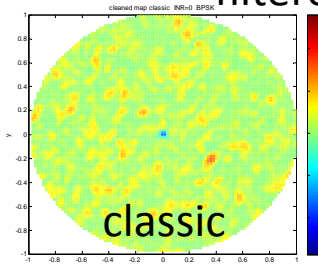


Simulations with 3 BPSK

Initial sky map



filtered sky maps



Cyclic Spatial Filtering (2)

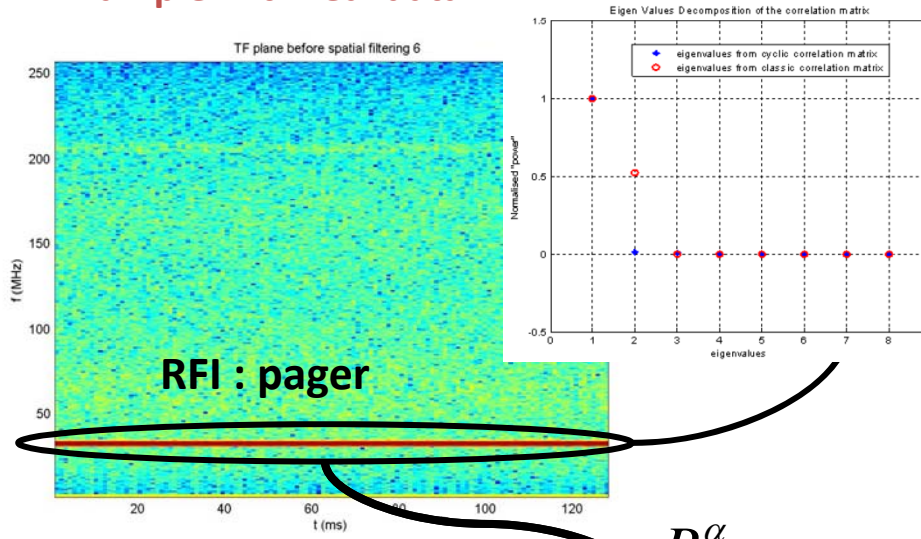


UNIVERSITÉ D'ORLÉANS

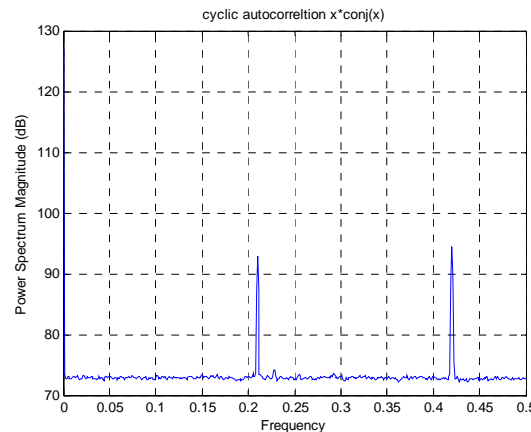
ASTRON

18

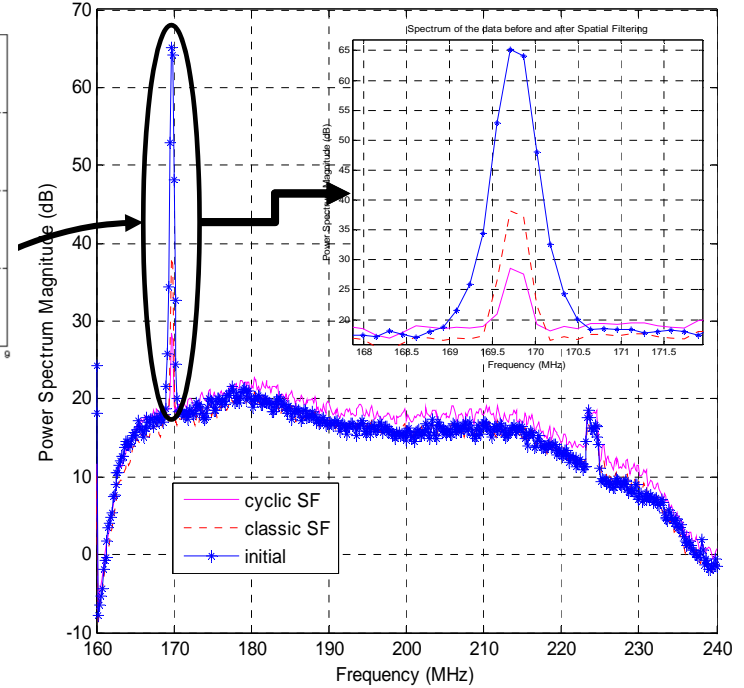
Example with real data from LOFAR



$$R^\alpha$$



Spectrum of the data before and after Spatial Filtering



Conclusions



Application		EC man month (mm) Deliverable : matlab golden model
Starting time		3 mm
Digital receiver		
	Impulse detector	3 mm
	Cyclostationary detector	3 mm
	Kurtosis detector	3 mm
	FFT2D + radon transform	4 mm
Beamformer		
	Pre -beamformer Cyclostationary detector	4mm
	Post-beamformer cyclostationary detector	3mm
	Spatial filtering	???
Correlator		
	Cyclostationary detector	6 mm
Pulsar machine		
	Upgrade of Uniboard design	3 mm

In collaboration with Astron, PhD student will start to work on spatial filtering