

Polynomial Matrix Decompositions Motivated by Defence Applications

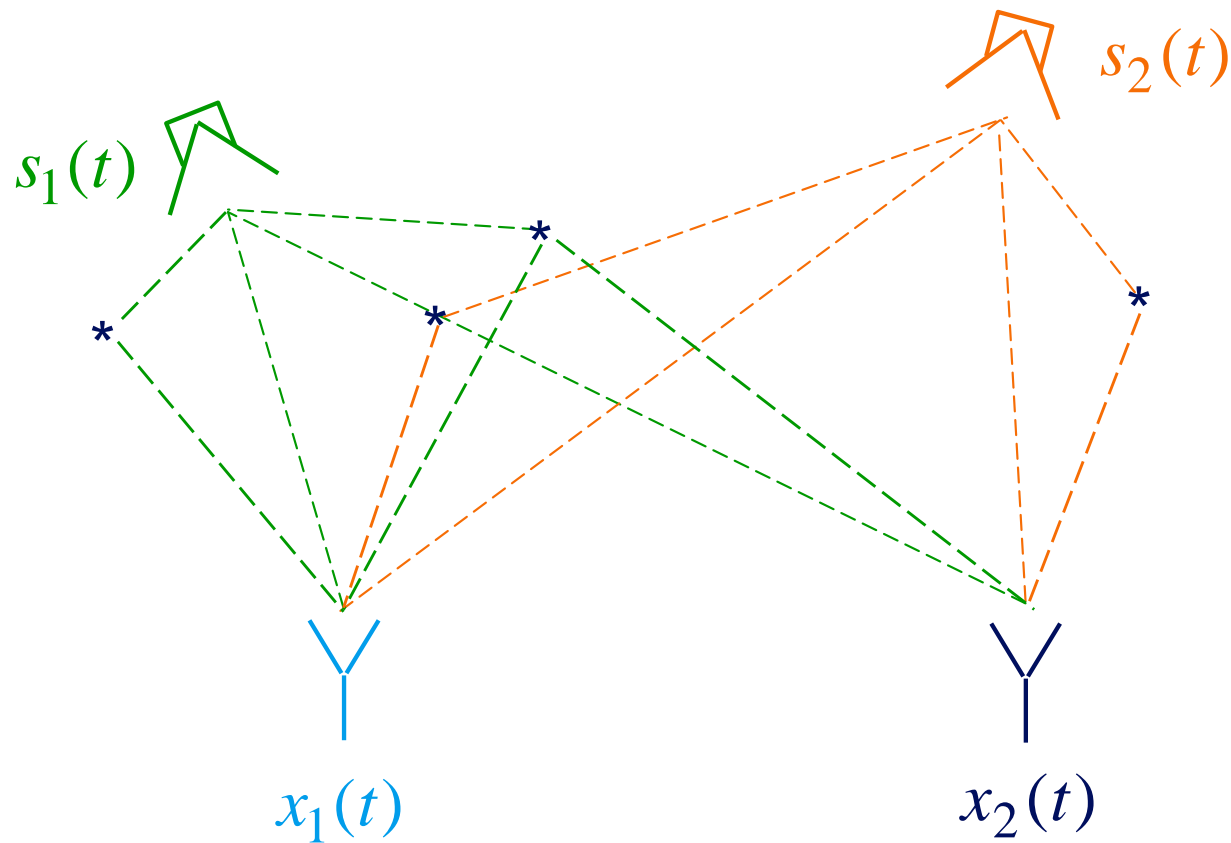
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Outline of Talk

- Convolutive mixing and polynomial matrices
 - strong decorrelation by PEVD
- Sequential best rotation algorithm (SBR2)
 - convolutive signal extraction
- Application to sonobuoy array processing
- Polynomial matrix QR decomposition (PQRD)
 - MIMO channel equalisation
- Other potential military applications
- Problem with coherent research funding

Convolutional Mixing

- Effects of multipath, dispersion etc



Channel Model

- Weighted sum of delayed samples (convolution)

$$x(n) = a_0s(n) + a_1s(n-1) + \dots\dots\dots a_p s(n-p)$$

- Express in *polynomial* form (c.f. z-transform)

$$\underline{a}(z) = a_0 + a_1z^{-1} + \dots\dots\dots a_p z^{-p}$$

$$\underline{s}(z) = s(0) + s(1)z^{-1} + \dots\dots\dots s(n)z^{-n} + \dots\dots\dots$$

$$\underline{x}(z) = x(0) + x(1)z^{-1} + \dots\dots\dots x(n)z^{-n} + \dots\dots\dots$$

- Convolution becomes *simple product*

$$\underline{x}(z) = \underline{a}(z)\underline{s}(z)$$

- Underscore used to denote polynomial quantities

Polynomial Matrix Formulation

- Two signals and two sensors

$$\underline{x}_1(z) = \underline{a}_{11}(z)\underline{s}_1(z) + \underline{a}_{12}(z)\underline{s}_2(z)$$

$$\underline{x}_2(z) = \underline{a}_{21}(z)\underline{s}_1(z) + \underline{a}_{22}(z)\underline{s}_2(z)$$

- Polynomial matrix form

$$\begin{bmatrix} \underline{x}_1(z) \\ \underline{x}_2(z) \end{bmatrix} = \begin{bmatrix} \underline{a}_{11}(z) & \underline{a}_{12}(z) \\ \underline{a}_{21}(z) & \underline{a}_{22}(z) \end{bmatrix} \begin{bmatrix} \underline{s}_1(z) \\ \underline{s}_2(z) \end{bmatrix}$$

- i.e

$$\underline{\mathbf{x}}(z) = \underline{\mathbf{A}}(z)\underline{\mathbf{s}}(z)$$

(Matrix) Polynomial (Matrix)

Diagram illustrating the structure of a (Matrix) Polynomial (Matrix) at three different time points: $-\tau$, 0 , and τ .

The matrices are arranged in a staircase pattern, showing the evolution of the polynomial matrix over time. The matrix at time 0 is highlighted in blue, while the matrices at $-\tau$ and τ are in grey.

The matrix at time $-\tau$ is:

$$\begin{bmatrix} r_{11}(-\tau) & r_{12}(-\tau) & r_{13}(-\tau) \\ r_{21}(-\tau) & r_{22}(-\tau) & r_{23}(-\tau) \\ r_{31}(-\tau) & r_{32}(-\tau) & r_{33}(-\tau) \end{bmatrix}$$

The matrix at time 0 is:

$$\begin{bmatrix} r_{11}(0) & r_{12}(0) & r_{13}(0) \\ r_{21}(0) & r_{22}(0) & r_{23}(0) \\ r_{32}(0) & r_{33}(0) \end{bmatrix}$$

The matrix at time τ is:

$$\begin{bmatrix} r_{11}(\tau) & r_{12}(\tau) & r_{13}(\tau) \\ r_{21}(\tau) & r_{22}(\tau) & r_{23}(\tau) \\ r_{32}(\tau) & r_{33}(\tau) \end{bmatrix}$$

Special Polynomial Matrices

- Unimodular Matrix $\det[\underline{\mathbf{A}}(z)] = \text{const}$
- Paraconjugation $\tilde{\underline{\mathbf{A}}}(z) = \underline{\mathbf{A}}_*^T(1/z)$
- Para-Hermitian Matrix $\tilde{\underline{\mathbf{A}}}(z) = \underline{\mathbf{A}}(z)$
- Paraunitary matrix (defines multichannel all-pass filter)

$$\underline{\mathbf{H}}(z)\tilde{\underline{\mathbf{H}}}(z) = \tilde{\underline{\mathbf{H}}}(z)\underline{\mathbf{H}}(z) = \mathbf{I}$$

Polynomial Matrix Examples

$$\underline{\mathbf{A}}(z) = \begin{bmatrix} 1 + z^{-1} & z + 2 \\ z^{-1} & 2 \end{bmatrix}$$

- Paraconjugate

$$\tilde{\underline{\mathbf{A}}}(z) = \begin{bmatrix} 1 + z & z \\ z^{-1} + 2 & 2 \end{bmatrix}$$

- Inverse

$$\underline{\mathbf{A}}^{-1}(z) = \begin{bmatrix} 2 & -(z + 2) \\ -z^{-1} & 1 + z^{-1} \end{bmatrix} \quad \det[\underline{\mathbf{A}}(z)] = 1$$

- Holds for all values of z - time domain processing

Polynomial Matrix EVD

$$\underline{\mathbf{R}}_{vv}(z) = \underline{\mathbf{H}}(z)\underline{\mathbf{R}}_{xx}(z)\tilde{\underline{\mathbf{H}}}(z) = \begin{bmatrix} \underline{d}_1(z) & 0 \\ 0 & \underline{d}_p(z) \end{bmatrix}$$

- $\underline{\mathbf{H}}(z)$ is a paraunitary; $\underline{\mathbf{R}}_{xx}(z)$ is para-Hermitian
- Taken as definition of polynomial matrix EVD (PEVD)
- For polynomials of order zero \rightarrow scalar EVD
- Corresponds to diagonalising cross spectral density matrix (Laurent polynomial) by paraunitary similarity transform

$$[\underline{\mathbf{R}}_{xx}(\tau)]_{ij} = \mathbb{E}\{x_i(t)x_j^*(t-\tau)\} \quad \underline{\mathbf{R}}_{xx}(z) = \sum_{\tau} \underline{\mathbf{R}}_{xx}(\tau)z^{-\tau}$$

Elementary Paraunitary Transformation

- Applies integer delay followed by scalar rotation

$$\underline{\Gamma}(z) = \begin{bmatrix} c & s \\ -s^* & c \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & z^{-N} \end{bmatrix}$$

- Note that

$$\underline{\Gamma}(z)\tilde{\underline{\Gamma}}(z) = \begin{bmatrix} c & s \\ -s^* & c \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & z^{-N} \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & z^N \end{bmatrix} \begin{bmatrix} c & -s \\ s^* & c \end{bmatrix} = \mathbf{I}$$

SBR2 Algorithm

- Given the estimated space-time covariance matrix

$$r_{ij}(\tau) = \sum_{t=1}^T x_i(t)x_j(t-\tau)/T$$

- Locate dominant cross-correlation coefficient $r_{kl}(t)$
- Shift to zero-lag position by applying relative delay operator
- Apply rotation to achieve decorrelation at zero lag

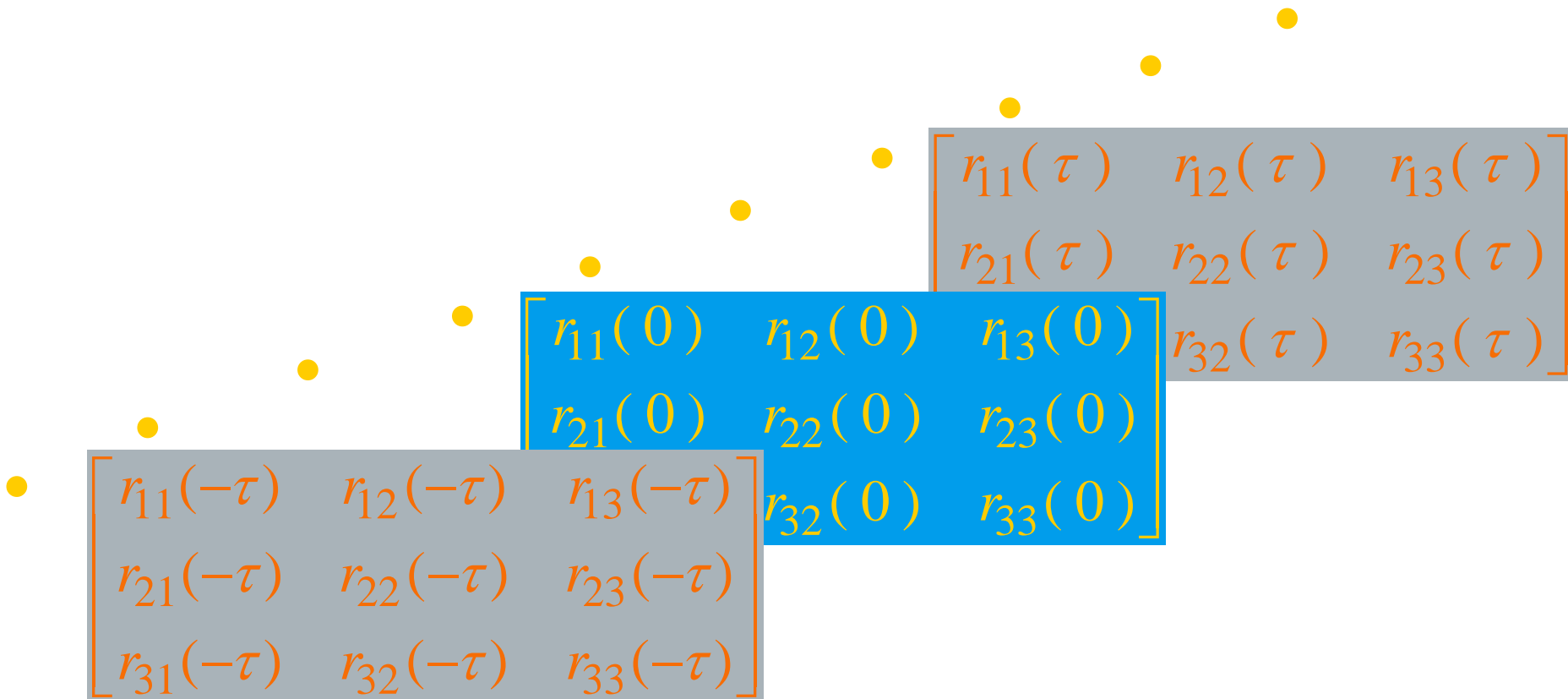
$$\tan(2\theta) = 2|r_{kl}(0)| / (r_{kk}(0) - r_{ll}(0))$$

$$r'_{kl}(0)$$

- Drives $r'_{kl}(0)$ to zero and updates covariance matrix
- Elementary paraunitary operation is repeated iteratively

SBR2 Algorithm

- Initial Polynomial covariance matrix



SBR2 Algorithm

- Single stage (assuming dominant coefficient is $r_{13}(1)$)
 - modifies every plane in polynomial matrix
 - order of polynomial grows (zero padding)

$$\begin{bmatrix} c & 0 & s \\ 0 & 1 & 0 \\ -s^* & 0 & c \end{bmatrix}$$

$$\begin{bmatrix} r_{11}(\tau) & r_{12}(\tau) & r_{13}(\tau+1) \\ r_{21}(\tau) & r_{22}(\tau) & r_{23}(\tau+1) \\ r_{32}(\tau-1) & r_{33}(\tau) & \end{bmatrix}$$

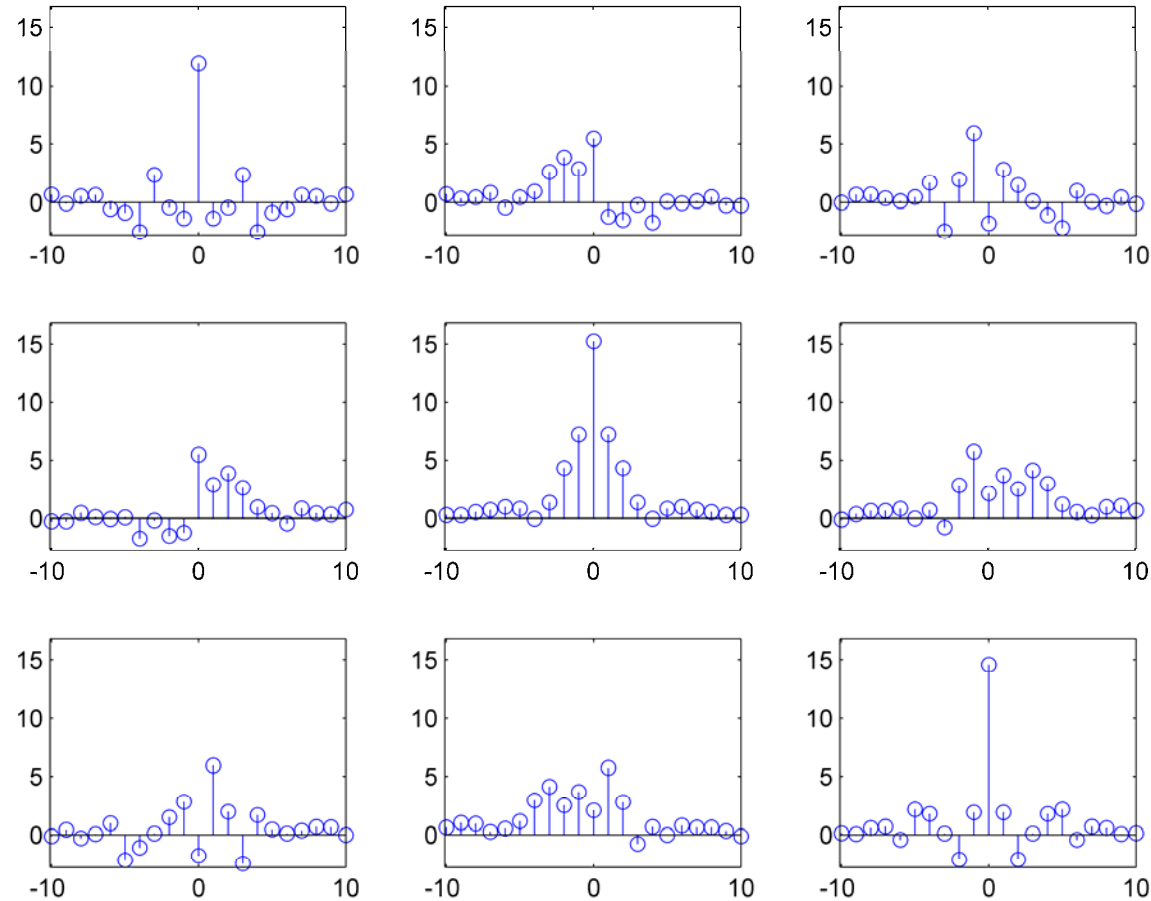
$$\begin{bmatrix} r_{11}(0) & r_{12}(0) & r_{13}(1) \\ r_{21}(0) & r_{22}(0) & r_{23}(1) \\ r_{33}(0) & & \end{bmatrix}$$

$$\begin{bmatrix} r_{11}(-\tau) & r_{12}(-\tau) & r_{13}(1-\tau) \\ r_{21}(-\tau) & r_{22}(-\tau) & r_{23}(1-\tau) \\ r_{31}(-1-\tau) & r_{32}(-1-\tau) & r_{33}(-\tau) \end{bmatrix}$$

$$\begin{bmatrix} c & 0 & -s \\ 0 & 1 & 0 \\ s^* & 0 & c \end{bmatrix}$$

Initial Polynomial Covariance Matrix

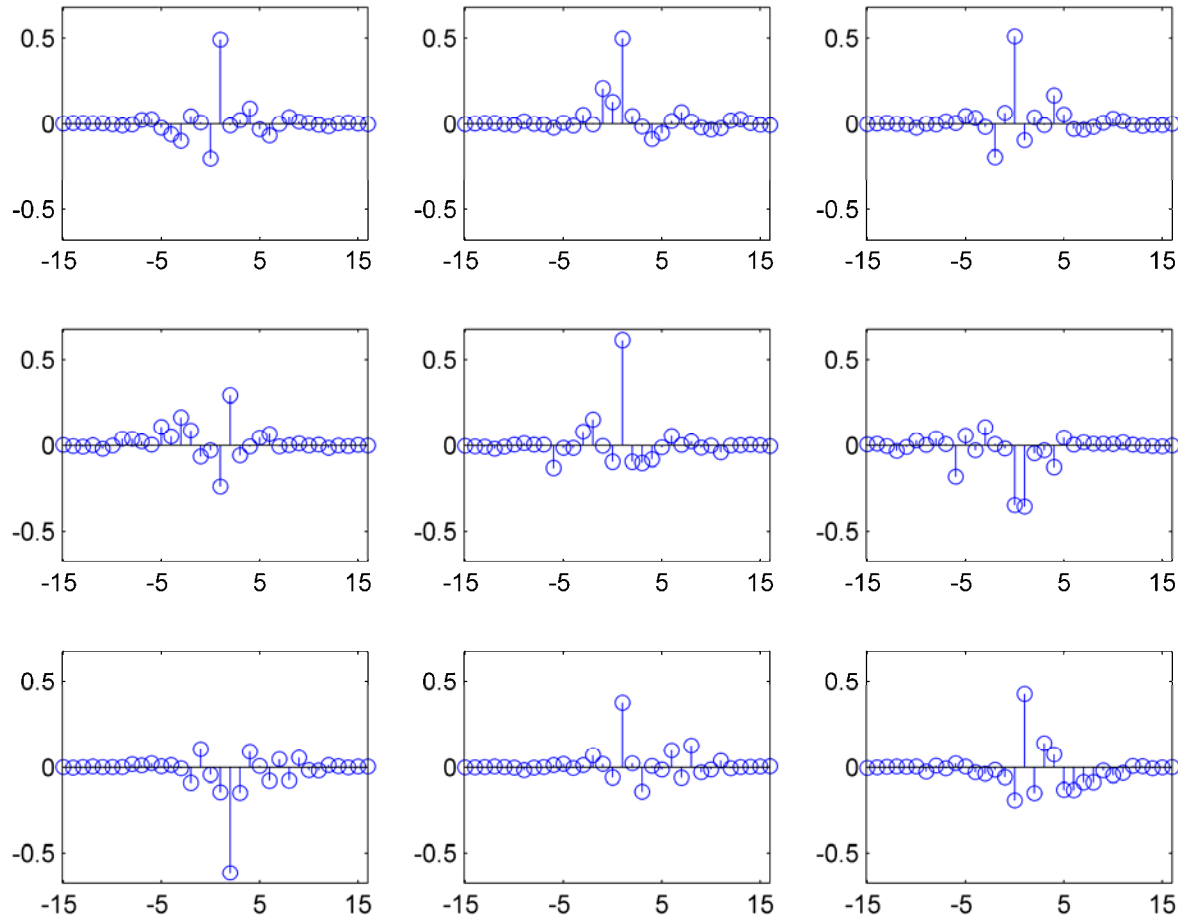
$\underline{R}_{xx}(z)$



2 sources, 3 sensors, order 5 mixing filters, coefficients random in [-1 1]

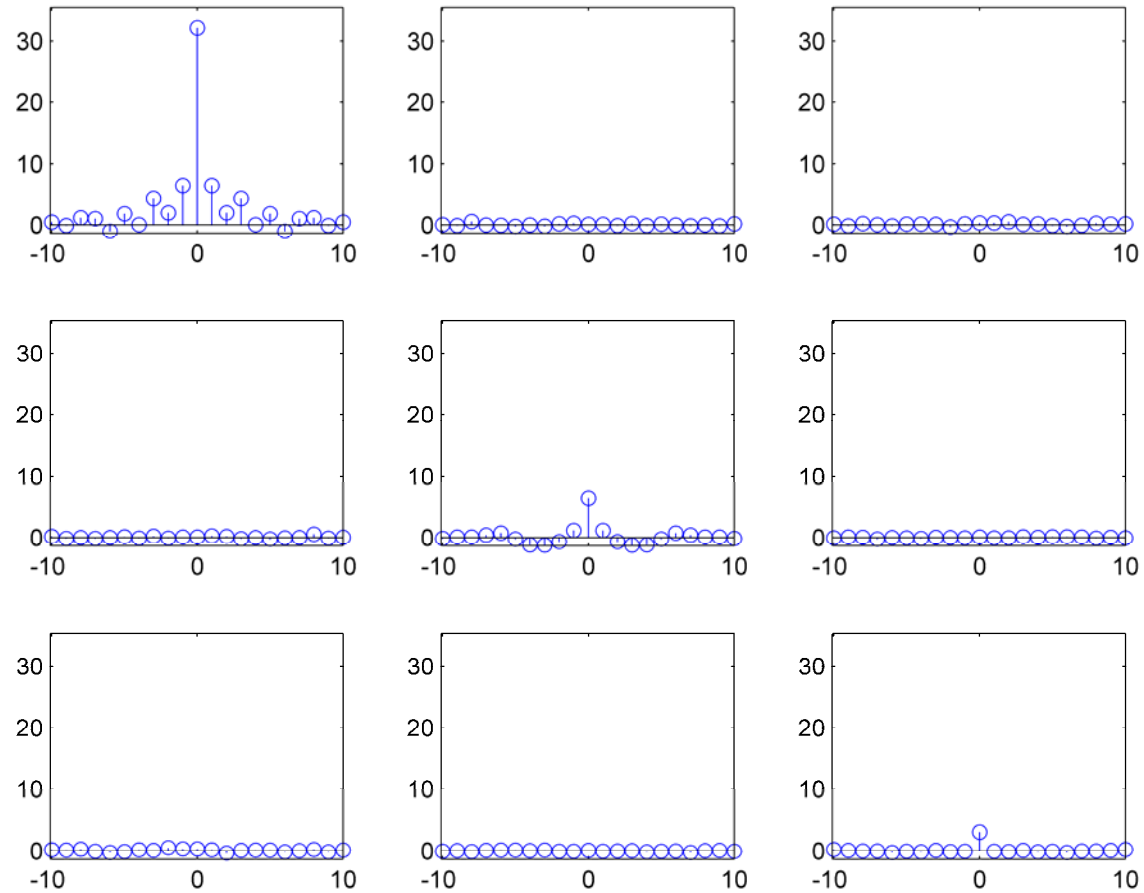
Paraunitary Transformation Matrix

$H(z)$

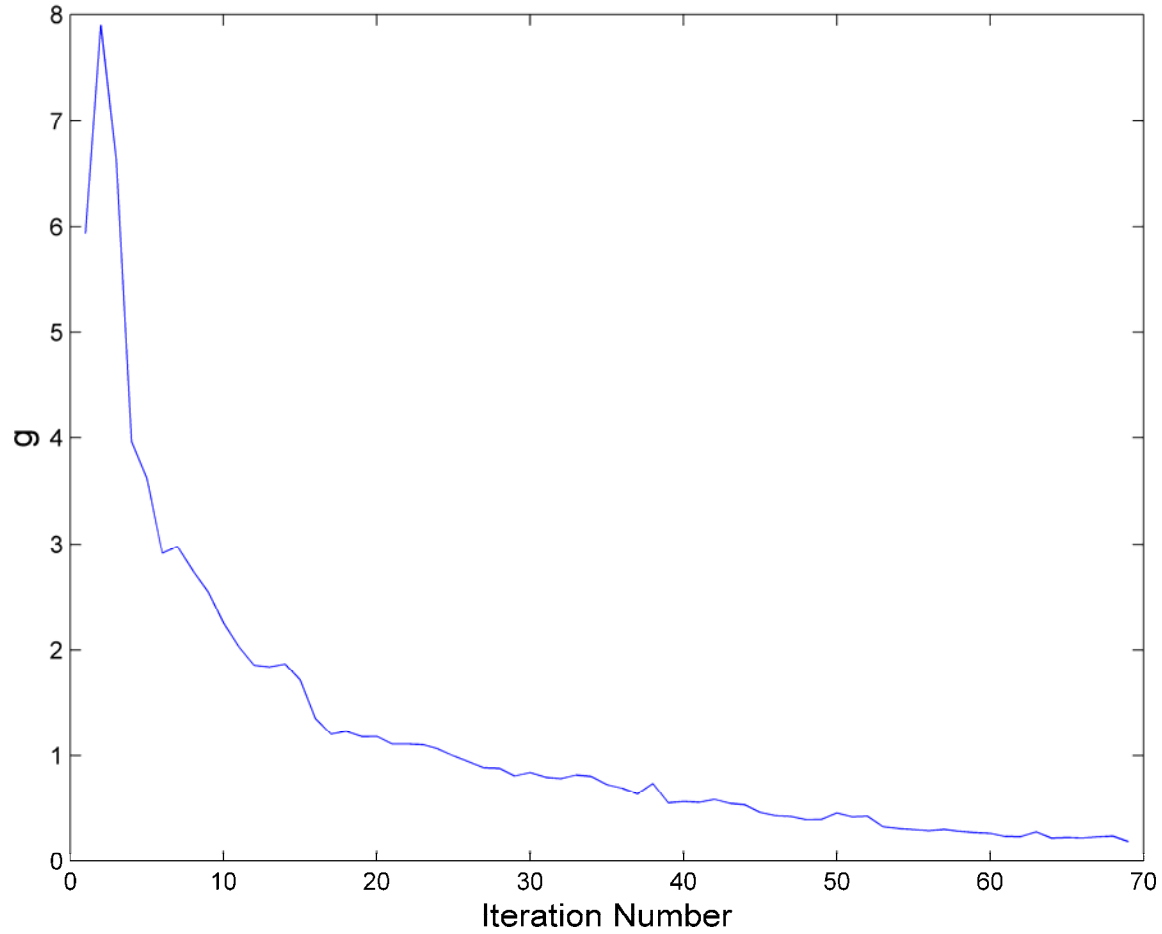


Effect of Strong Decorrelation

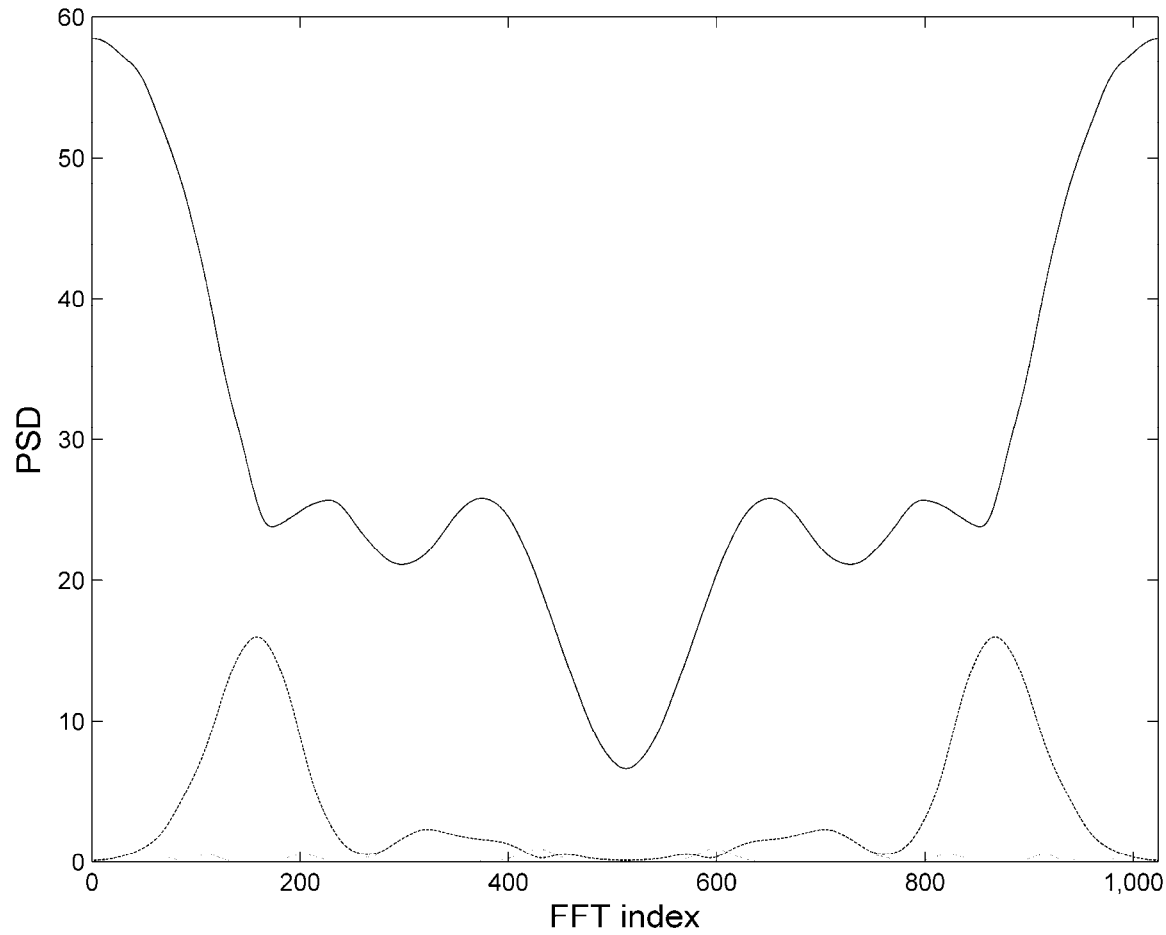
$\underline{R}_{vv}(z)$



Convergence of SBR2 Algorithm



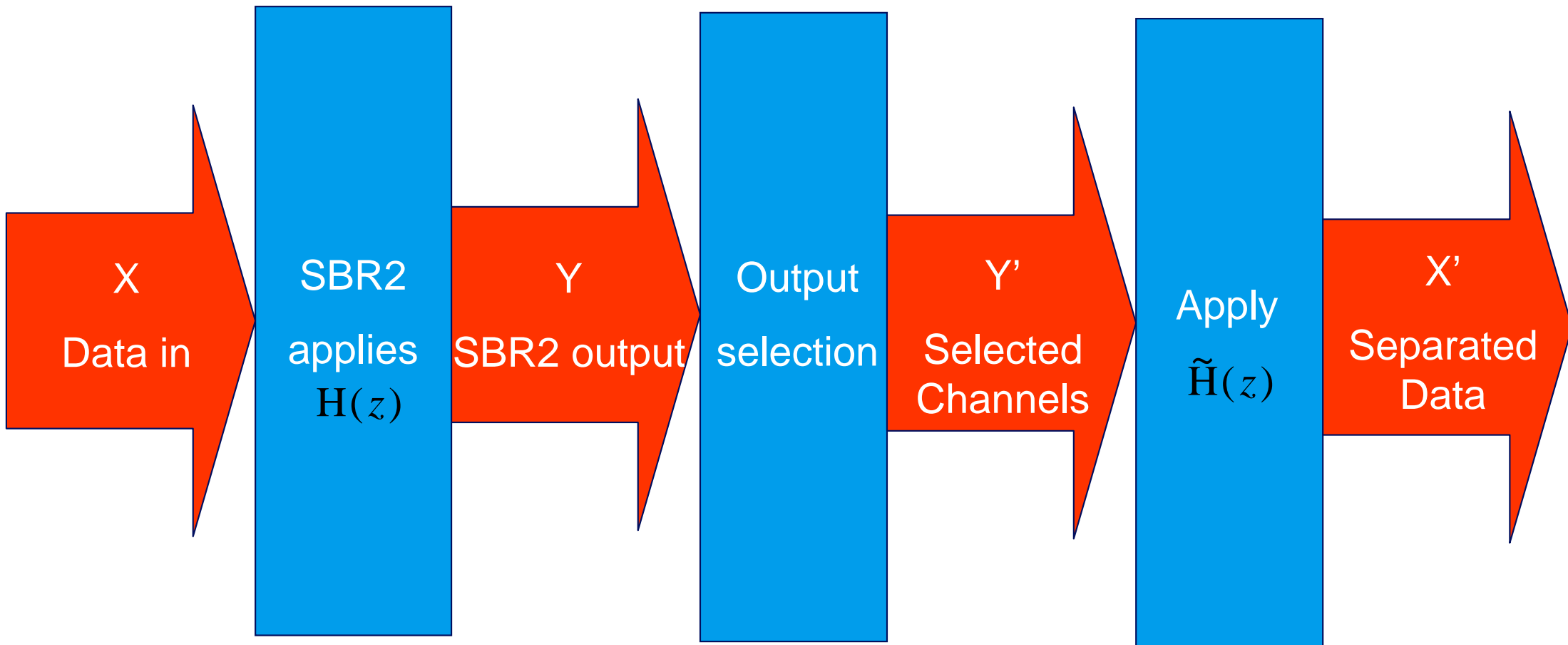
Spectral Majorisation of Output Signals



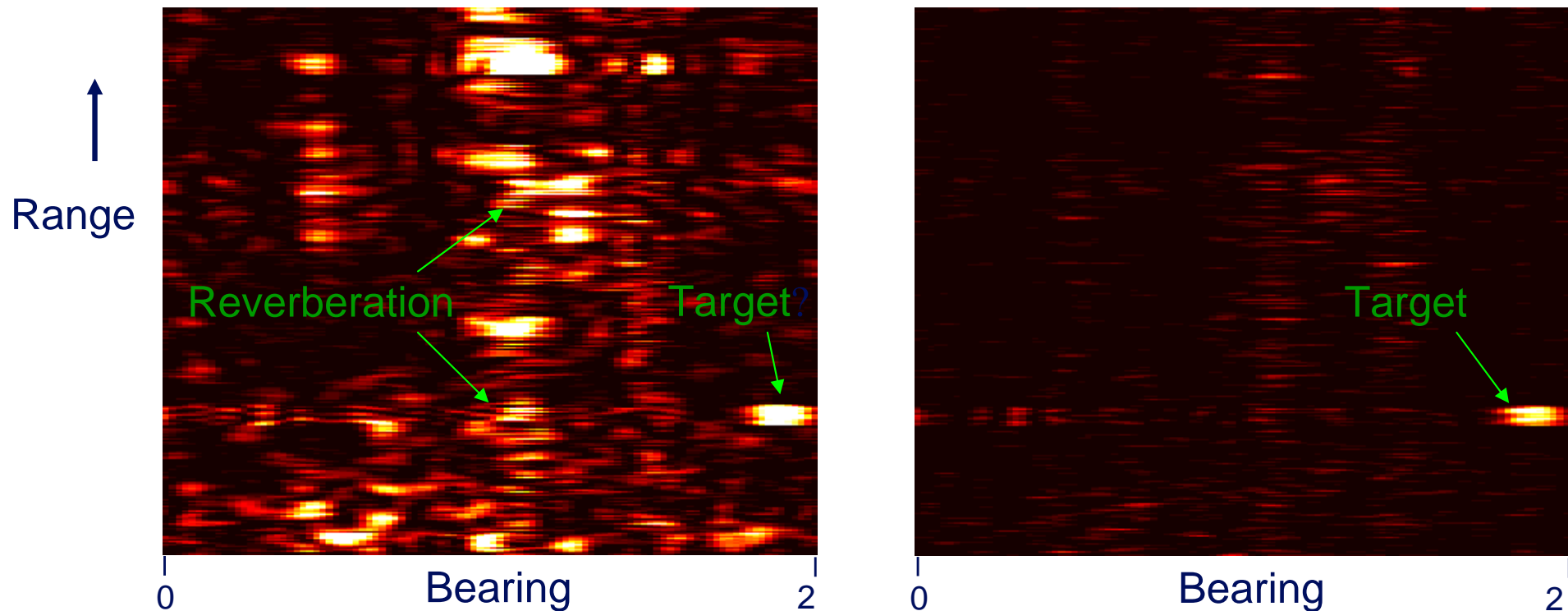
Sonobuoy Array



Signal separation using SBR2



Sonar reverberation suppression



- The SBR2 algorithm can be used to reduce reverberation considerably
- A target is revealed at a bearing of $\sim 343^\circ$

QR Decomposition of a Polynomial Matrix

Input matrix $\underline{\mathbf{A}}(z) \in \underline{\mathbb{C}}^{p \times q}$

Objective is to calculate $\underline{\mathbf{Q}}(z) \in \underline{\mathbb{C}}^{p \times p}$ such that

$$\underline{\mathbf{A}}(z) = \underline{\mathbf{Q}}(z)\underline{\mathbf{R}}(z)$$

where $\underline{\mathbf{R}}(z) \in \underline{\mathbb{C}}^{p \times q}$ is upper triangular and $\underline{\mathbf{Q}}(z)$ is paraunitary, i.e.

$$\underline{\tilde{\mathbf{Q}}}(z)\underline{\mathbf{Q}}(z) = \underline{\mathbf{Q}}(z)\underline{\tilde{\mathbf{Q}}}(z) = \mathbf{I}$$

MIMO Channel Equalisation

Convolutional Mixing Model

$$\mathbf{x}(t) = \sum_{k=0}^N \mathbf{A}(k) \mathbf{s}(t-k) + \mathbf{n}(t)$$

Or expressed in polynomial form

$$\underline{\mathbf{x}}(z) = \underline{\mathbf{A}}(z) \underline{\mathbf{s}}(z) + \underline{\mathbf{n}}(z)$$

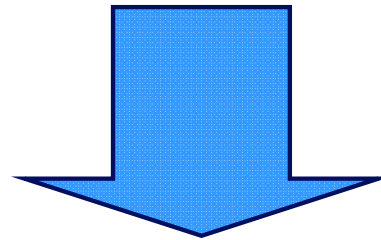
Assume channel matrix is known, then $\underline{\mathbf{A}}(z) = \underline{\mathbf{Q}}(z) \underline{\mathbf{R}}(z)$

Rearranging

$$\underbrace{\underline{\tilde{\mathbf{Q}}}(z) \underline{\mathbf{x}}(z)}_{\underline{\mathbf{x}}'(z)} = \underline{\mathbf{R}}(z) \underline{\mathbf{s}}(z) + \underbrace{\underline{\tilde{\mathbf{Q}}}(z) \underline{\mathbf{n}}(z)}_{\underline{\mathbf{n}}'(z)}$$

2x2 Example

$$\begin{bmatrix} \underline{x}'_1(z) \\ \underline{x}'_2(z) \end{bmatrix} = \begin{bmatrix} \underline{r}_{11}(z) & \underline{r}_{12}(z) \\ 0 & \underline{r}_{22}(z) \end{bmatrix} \begin{bmatrix} \underline{s}_1(z) \\ \underline{s}_2(z) \end{bmatrix} + \begin{bmatrix} \underline{n}'_1(z) \\ \underline{n}'_2(z) \end{bmatrix}$$



Use **back substitution** to...

Single channel
equalisation
problem – solve
using a maximum
likelihood
sequence estimator

1. Estimate source 2

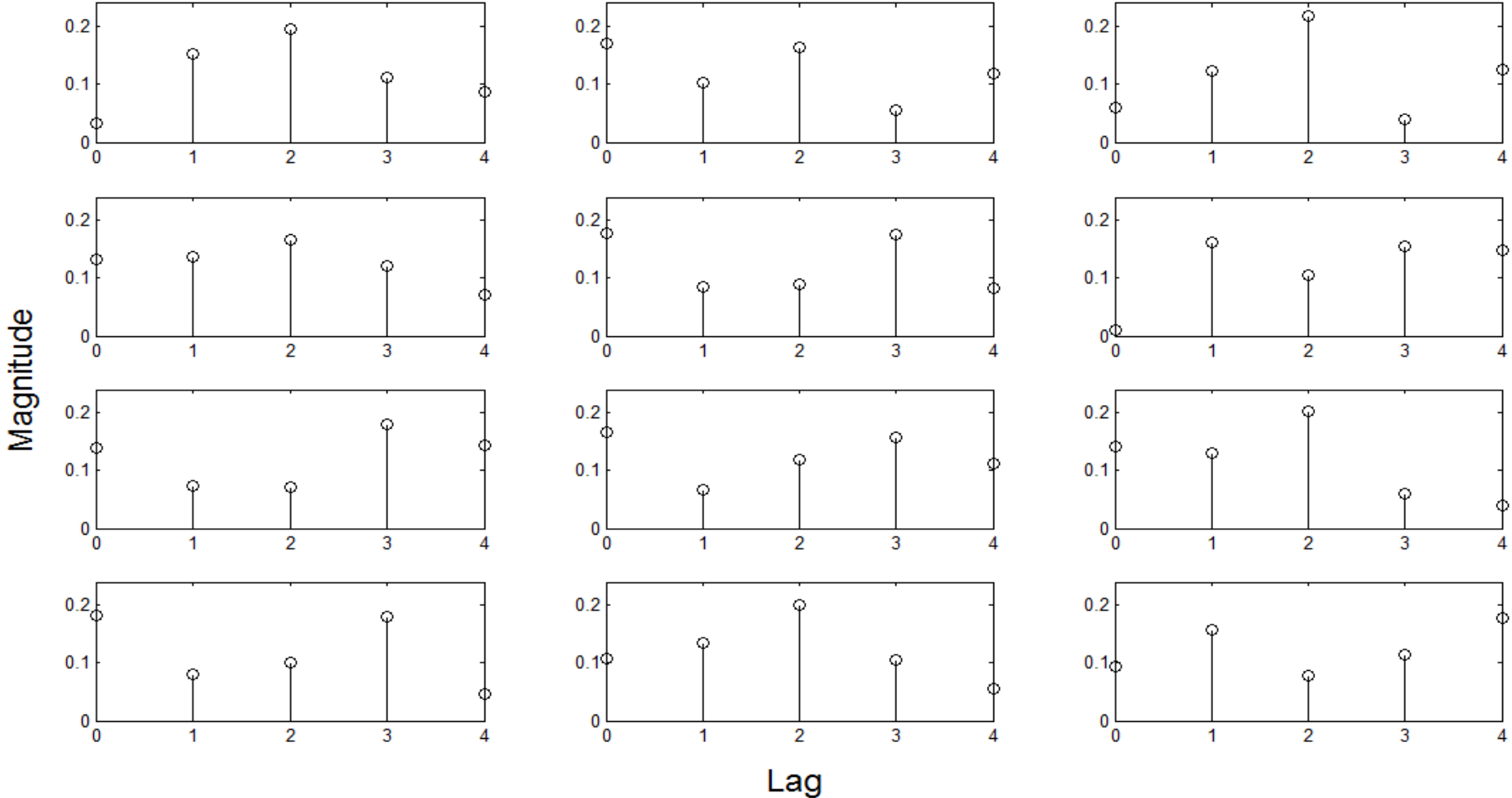
$$\underline{x}'_2(z) = \underline{r}_{22}(z)\underline{s}_2(z) + \underline{n}'_2(z)$$

2. Estimate source 1

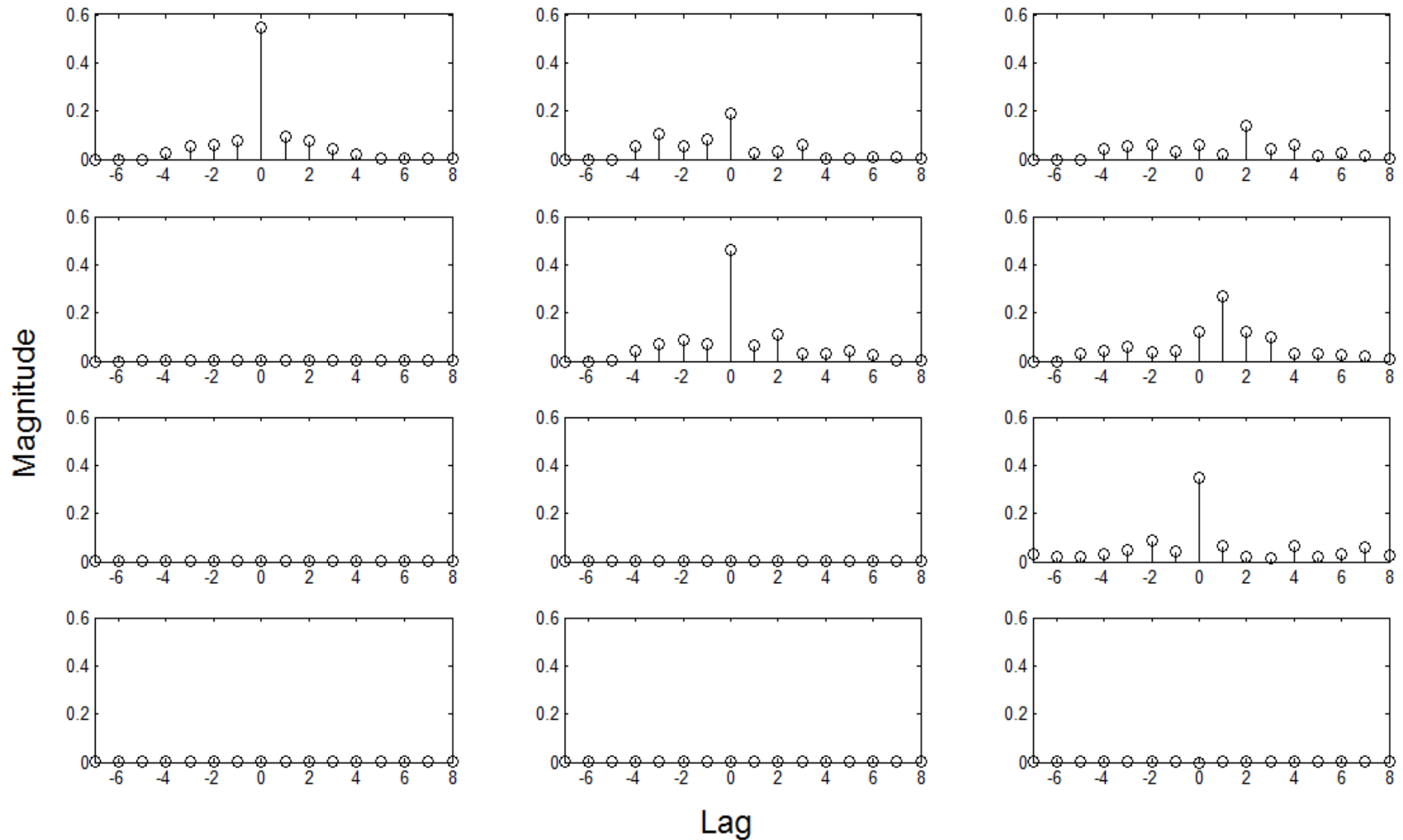
$$\underline{x}'_1(z) - \underline{r}_{12}(z)\underline{s}_2(z) = \underline{r}_{11}(z)\underline{s}_1(z) + \underline{n}'_1(z)$$

Simple Example

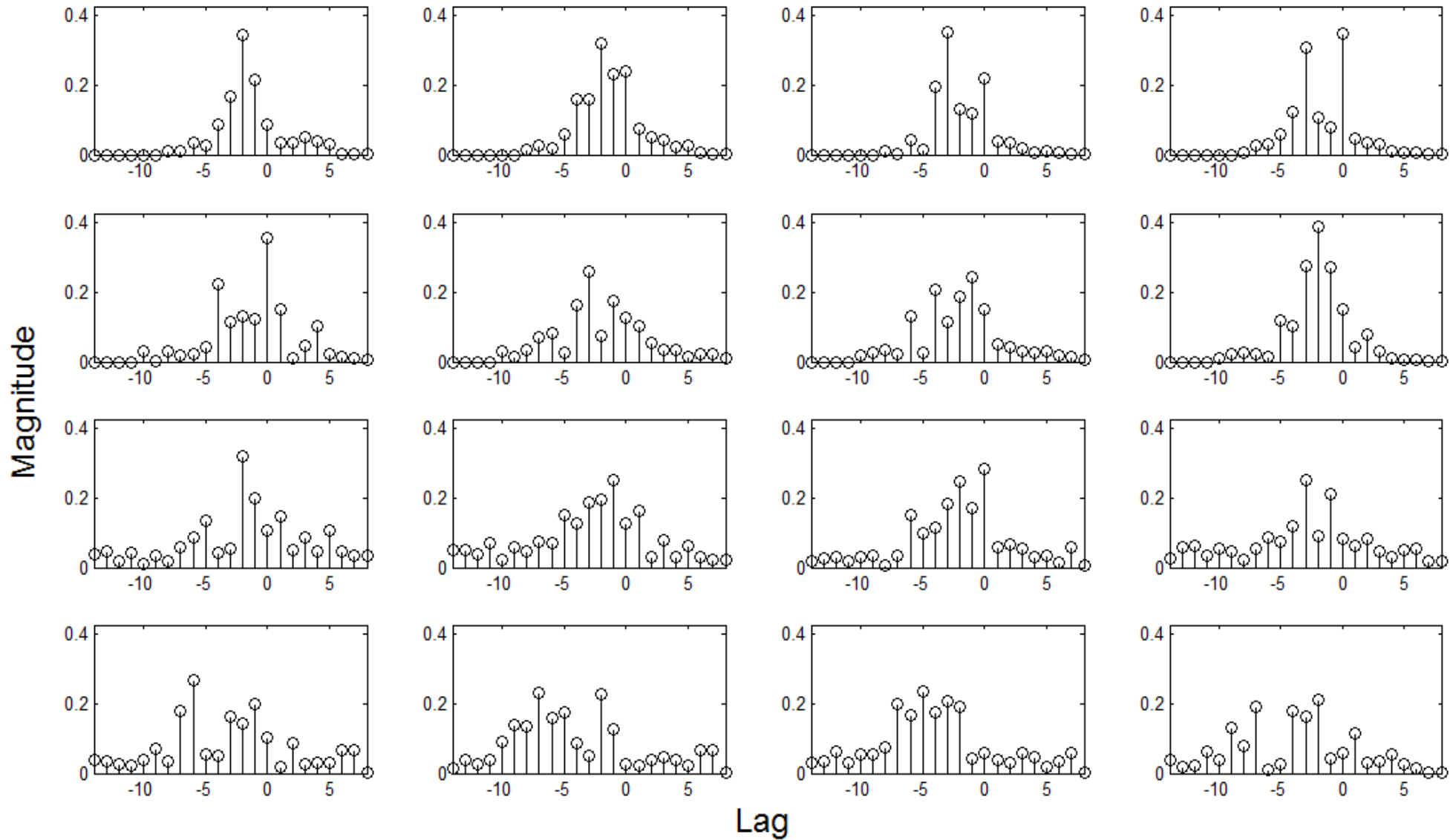
Input Matrix $\underline{A}(z)$



Upper Triangular Matrix $\underline{R}(z)$



Paraunitary Matrix $Q(z)$



Potential Military Applications

- Sonobuoy array processing
- Sonar towed array processing
- Sonar flank array processing
- Underwater acoustic communications
- STAP radar
- MIMO communication networks
- Monitoring seismic events (nuclear test ban treaty)
- Acoustic monitoring for ground asset or harbour protection

Important aspects to note

- Mathematical research motivated by military problems
- Initial research produces some “practical” algorithms
- Early simulations demonstrate potential of new approach
- Need for more fundamental mathematical analysis
- Opens up possibility of whole new research topic
 - Other numerical algorithms and approaches?
 - Stronger convergence and approximation results?
- Numerous applications to be investigated
- But still very difficult to obtain coherent funding

Acknowledgements

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- Reference
 - J G McWhirter, P D Baxter, T Cooper, S Redif and J Foster
“An EVD Algorithm for Para-Hermitian Polynomial Matrices”
IEEE Trans Signal Processing, Vol 55, No 6 (May 2007).