
Novel Quaternion Matrix Factorisations



Imperial College
London

S. Enshaeifar, C. Cheong Took, S. Sanei, and D. P. Mandic*

Faculty of Engineering and Physical Sciences,
University of Surrey, Guildford, U.K.

*Department of Electrical and Electronic Engineering,
Imperial College London, London, U.K.

Outline

- The big picture
- Simultaneous diagonalisation
 - Useful properties
 - Simultaneous diagonalisations in \mathbb{H}
 - Quaternion uncorrelating transform
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 - Alamouti-based communication problem
- Conclusion

The big picture

Quaternions are of interest in

- Computer Science
- Aeronautics
- Mathematics

Analysis in \mathbb{C} and \mathbb{H}

- Analysis of phase information $x = |x|e^{\phi}$
- Complex-valued problems in communication
- Diagonalisation of covariance matrices: a widely used procedure in signal processing algorithms such as PCA and ICA

Good to know: joint diagonalisation in \mathbb{C}

- Advances in \mathbb{C} : the necessity of ‘augmented statistics’ to incorporate the covariance and pseudo-covariance matrices:

$$\mathbf{z} = \mathbf{z}_r + j\mathbf{z}_i \in \mathbb{C} \rightarrow \begin{cases} \Re\{\mathbf{z}\} = \frac{1}{2}(\mathbf{z} + \mathbf{z}^*) \\ \Im\{\mathbf{z}\} = \frac{1}{2j}(\mathbf{z} - \mathbf{z}^*) \end{cases}$$

$$\mathbf{C}_z = E[\mathbf{z}\mathbf{z}^H] \quad \mathbf{P}_z = E[\mathbf{z}\mathbf{z}^T]$$

$$\mathring{\mathbf{z}} = [\mathbf{z}, \mathbf{z}^*]^T \rightarrow \mathbf{C}_{\mathring{z}} = \begin{bmatrix} \mathbf{C}_z & \mathbf{P}_z \\ \mathbf{P}_z^* & \mathbf{C}_z^* \end{bmatrix}$$

Strong uncorrelating transform (SUT): joint diagonalisation of \mathbf{C}_z and \mathbf{P}_z allows for the augmented signal processing algorithms in \mathbb{C} , known as SUT.

Useful properties

- **Quaternion involution** $(\cdot)^\eta$: an important notion for \mathbb{H}

$$\mathbf{q} = \mathbf{q}_r + i\mathbf{q}_i + j\mathbf{q}_j + k\mathbf{q}_k \in \mathbb{H} \rightarrow \begin{cases} \Re\{\mathbf{q}\} = \frac{1}{2}(\mathbf{q} + \mathbf{q}^*) \\ \Im_\eta\{\mathbf{q}\} = \frac{1}{2\eta}(\mathbf{q} - \mathbf{q}^{\eta*}) \end{cases}$$

$$\mathbf{q}^\eta = -\eta\mathbf{q}\eta, \quad \text{e.g.} \quad \mathbf{q}^i = -i\mathbf{q}i = \mathbf{q}_r + i\mathbf{q}_i - j\mathbf{q}_j - k\mathbf{q}_k$$

- Augmented statistics are also essential for \mathbb{H} to incorporate the covariance and η -covariance matrices:

$$\mathbf{C}_{\mathbf{q}} = E[\mathbf{q}\mathbf{q}^H] \quad \mathbf{C}_{\mathbf{q}^\eta} = E[\mathbf{q}\mathbf{q}^{\eta H}]$$

$$\mathring{\mathbf{q}} = [\mathbf{q}, \mathbf{q}^i, \mathbf{q}^j, \mathbf{q}^k]^T$$

Quaternion uncorrelating transform (QUT): simultaneous diagonalisation of $\mathbf{C}_{\mathbf{q}}$ and $\mathbf{C}_{\mathbf{q}^\eta}$ represents quaternion uncorrelating transform (QUT).

Useful properties

(1) A Hermitian matrix: $\mathbf{A} = \mathbf{A}^H \rightarrow$ standard covariance matrix

(2) An η -Hermitian matrix: $\mathbf{A} = \mathbf{A}^{\eta H} \rightarrow \eta$ -covariance matrices

$$\mathbf{C}_{\mathbf{x}\alpha} = E\{\mathbf{x}\mathbf{x}^{\alpha H}\} \quad \alpha \in \{\iota, \jmath, \kappa\} \quad (1)$$

$$\mathbf{x}\mathbf{x}^{\alpha H} = \begin{bmatrix} x_1 x_1^{\alpha*} & x_1 x_2^{\alpha*} & \cdots & x_1 x_N^{\alpha*} \\ x_2 x_1^{\alpha*} & x_2 x_2^{\alpha*} & \cdots & x_2 x_N^{\alpha*} \\ \vdots & \vdots & \ddots & \vdots \\ x_N x_1^{\alpha*} & x_N x_2^{\alpha*} & \cdots & x_N x_N^{\alpha*} \end{bmatrix}$$

Quaternion Takagi factorisation¹ for η -Hermitian matrix \mathbf{A}

$$\mathbf{A} = \mathbf{Q}\mathbf{S}_\eta\mathbf{Q}^{\eta H} \begin{cases} \mathbf{Q} \text{ is a quaternion unitary matrix} \\ \mathbf{S}_\eta \text{ is a real-valued non-negative diagonal matrix} \end{cases}$$

¹MATLAB code available at http://www.surrey.ac.uk/cs/people/clive_cheong_took/index.htm

Simultaneous diagonalisations in \mathbb{H} - case 1

Introduction of η -Hermitian: three sets of joint diagonalisation for matrices $\mathbf{A}, \mathbf{B} \in \mathbb{H}$:

(1) If \mathbf{A} and \mathbf{B} are both Hermitian, there exists a matrix $\mathbf{M} \in \mathbb{H}$ such that $\mathbf{M}^H \mathbf{A} \mathbf{M}$ and $\mathbf{M}^H \mathbf{B} \mathbf{M}$ are both diagonal if and only if $\mathbf{A} \mathbf{B}$ is Hermitian, i.e. $\mathbf{A} \mathbf{B} = \mathbf{B} \mathbf{A}$

Since $\mathbf{A} = \mathbf{U} \mathbf{S}_a \mathbf{U}^H$ and \mathbf{B} are both Hermitian:

$$\mathbf{D} = \mathbf{U} \mathbf{S}_a^{-\frac{1}{2}} \mathbf{U}^H \quad \text{so} \quad \left\{ \begin{array}{l} \mathbf{D} \mathbf{A} \mathbf{D}^H = \mathbf{I} \\ \mathbf{D} \mathbf{B} \mathbf{D}^H = \mathbf{W} \mathbf{\Lambda}_b \mathbf{W}^H \end{array} \right\}$$

consider $\mathbf{M} = \mathbf{D}^H \mathbf{W}$, thus:

$$\mathbf{M}^H \mathbf{A} \mathbf{M} = \mathbf{W}^H \mathbf{D} \mathbf{A} \mathbf{D}^H \mathbf{W} = \mathbf{I} = \mathbf{\Lambda}_a$$

$$\mathbf{M}^H \mathbf{B} \mathbf{M} = \mathbf{W}^H (\mathbf{D} \mathbf{B} \mathbf{D}^H) \mathbf{W} = \mathbf{\Lambda}_b$$

Simultaneous diagonalisations in \mathbb{H} - case 2

(2) If \mathbf{A} and \mathbf{B} are both η -Hermitian, there exists a unitary matrix $\mathbf{M} \in \mathbb{H}$ such that $\mathbf{M}^{\eta H} \mathbf{A} \mathbf{M}$ and $\mathbf{M}^{\eta H} \mathbf{B} \mathbf{M}$ are both diagonal if and only if $\mathbf{A} \mathbf{B}^\eta$ is normal, i.e. $(\mathbf{A} \mathbf{B}^\eta)(\mathbf{A} \mathbf{B}^\eta)^H = (\mathbf{A} \mathbf{B}^\eta)^H (\mathbf{A} \mathbf{B}^\eta)$

A single unitary matrix \mathbf{M} is sufficient to diagonalise \mathbf{A} and \mathbf{B} simultaneously.

$$\mathbf{A} \stackrel{\text{Q-SVD}}{=} \mathbf{U} \mathbf{S} \mathbf{V}^H$$

$$\mathbf{A} \stackrel{\text{Q-Takagi}}{=} \mathbf{Q} \mathbf{S} \mathbf{Q}^{\eta H} \quad \text{where} \quad \begin{cases} \mathbf{D} = \mathbf{V}^{\eta H} \mathbf{U} \\ \mathbf{Q} = \mathbf{U} (\mathbf{D}^\eta)^{\frac{1}{2}} \end{cases}$$

Consider $\mathbf{M} = \mathbf{Q}^\eta$, thus:

$$\mathbf{M}^{\eta H} \mathbf{A} \mathbf{M} = \mathbf{Q}^H \mathbf{A} \mathbf{Q}^\eta = \mathbf{Q}^H (\mathbf{Q} \mathbf{S} \mathbf{Q}^{\eta H}) \mathbf{Q}^\eta = \mathbf{S}$$

diagonality of $\mathbf{M}^{\eta H} \mathbf{B} \mathbf{M}$ can be proved using the normality of $\mathbf{A} \mathbf{B}^\eta$

Simultaneous diagonalisations in \mathbb{H} - case 3

(3) If \mathbf{A} is Hermitian and \mathbf{B} is η -Hermitian, there exists a matrix $\mathbf{M} \in \mathbb{H}$ such that $\mathbf{M}^H \mathbf{A} \mathbf{M}$ and $\mathbf{M}^{\eta H} \mathbf{B} \mathbf{M}$ are both diagonal if and only if $\mathbf{B} \mathbf{A}$ is η -Hermitian, i.e. $\mathbf{B} \mathbf{A} = (\mathbf{B} \mathbf{A})^{\eta H} = \mathbf{A}^{\eta} \mathbf{B}$

Since $\mathbf{A} = \mathbf{U} \mathbf{S}_a \mathbf{U}^H$ is Hermitian and \mathbf{B} is η -Hermitian:

$$\mathbf{D} = \mathbf{U} \mathbf{S}_a^{-\frac{1}{2}} \mathbf{U}^H \quad \text{so} \quad \left\{ \begin{array}{l} \mathbf{D} \mathbf{A} \mathbf{D}^H = \mathbf{I} \\ \mathbf{D}^{\eta} \mathbf{B} (\mathbf{D}^{\eta})^{\eta H} = \mathbf{W} \mathbf{\Lambda}_b \mathbf{W}^{\eta H} \end{array} \right\}$$

Consider $\mathbf{M} = \mathbf{D}^H \mathbf{W}^{\eta}$, thus:

$$\begin{aligned} \mathbf{M}^H \mathbf{A} \mathbf{M} &= \mathbf{W}^{\eta H} \mathbf{D} \mathbf{A} \mathbf{D}^H \mathbf{W}^{\eta} = \mathbf{I} = \mathbf{\Lambda}_a \\ \mathbf{M}^{\eta H} \mathbf{B} \mathbf{M} &= \mathbf{W}^H (\mathbf{D}^{\eta} \mathbf{B} \mathbf{D}^{\eta H}) \mathbf{W}^{\eta} = \mathbf{\Lambda}_b \end{aligned}$$

Generalisation of case 3: quaternion uncorrelating transform

- **Assumptions in case 3:** \mathbf{A} is Hermitian and \mathbf{B} is η -Hermitian
- **Results of case 3:** $\mathbf{M}^H \mathbf{A} \mathbf{M} = \mathbf{I}$ and $\mathbf{M}^{\eta H} \mathbf{B} \mathbf{M} = \mathbf{\Lambda}_b$
- **Remark:** In general, covariance matrix is Hermitian and η -covariance matrix is η -Hermitian



Quaternion uncorrelating transform (QUT): for a random quaternion vector \mathbf{x} , there exists a QUT matrix \mathbf{M} which simultaneously whitens the covariance and diagonalises an η -covariance matrix of $\mathbf{y} = \mathbf{M}^H \mathbf{x}$, i.e. $\mathbf{C}_y = \mathbf{I}$ and $\mathbf{C}_{y\eta} = \mathbf{\Lambda}_\eta$.

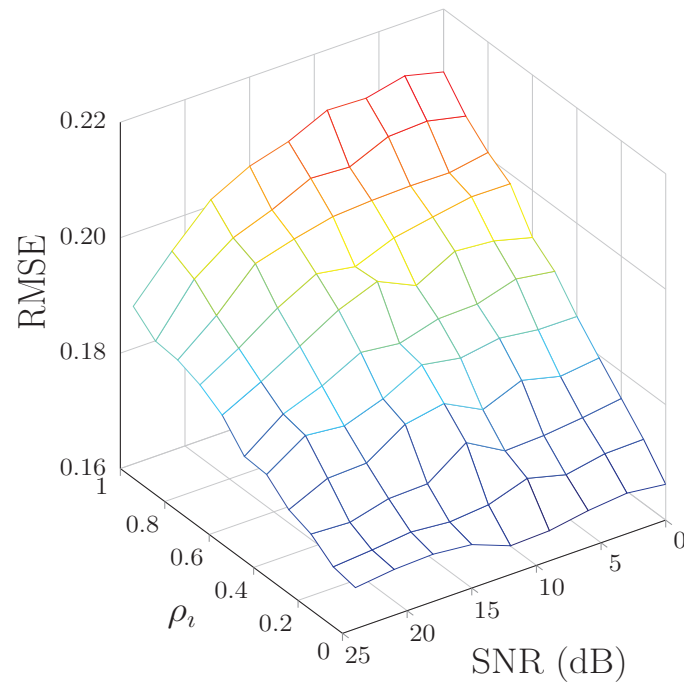
Properness and quaternion strong uncorrelating transform

- **Properness:** an important statistical property in both \mathbb{C} and \mathbb{H} .
- Characterised by the degree of correlation and/or power difference
- Properness depends on the pseudo-covariance in \mathbb{C}
- **Two types of properness in \mathbb{H} :**
 - * \mathbb{H} -proper: if and only if all the η -covariance matrices vanish.
 - * \mathbb{C}^η -proper for a single imaginary unit $\eta \in \{i, j, \kappa\}$: if only the η -covariance matrix exists and the other two vanish.
 - \mathbf{x} and \mathbf{x}^η are correlated through the $E[\mathbf{x}(\mathbf{x}^\eta)^H]$
 - The degree of \mathbb{C} -properness for a quaternion variable x is:
$$\rho = \frac{|E\{xx^{\eta H}\}|}{|E\{xx^H\}|} \quad \rho \in [0, 1]$$
- For a \mathbb{C}^η -proper data, QUT can be considered as the quaternion strong uncorrelating transform (Q-SUT) which diagonalises the covariance and all three η -covariance matrices.

Simulation 1

Evaluate the performance accuracy in terms of

1. The additive white noise, SNR values
2. The ν -circularity coefficient



Simulation 2: Alamouti-based coding

Assess the proposed QUT for a practical communication problem of Alamouti coding

- For a single-user, the model of two transmit antennas (a, b) and one receiver antenna:

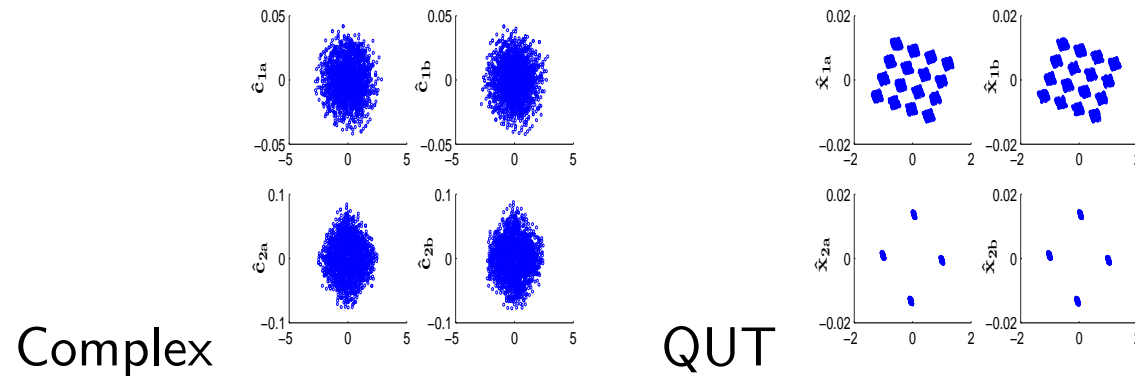
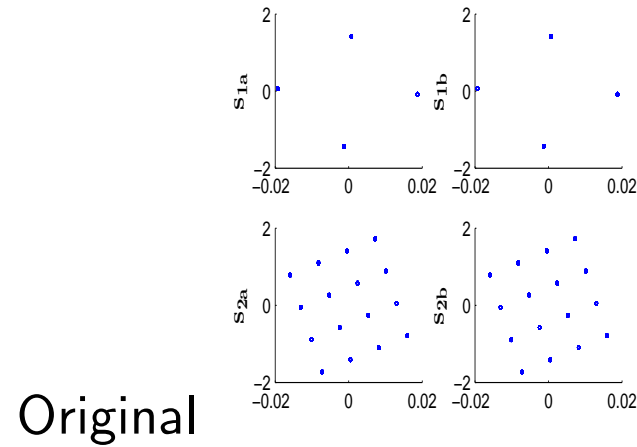
$$\begin{bmatrix} \mathbf{x}_a \\ \mathbf{x}_b \end{bmatrix} = \begin{bmatrix} \mathbf{s}_a & -\mathbf{s}_b^* \\ \mathbf{s}_b & \mathbf{s}_a^* \end{bmatrix} \begin{bmatrix} \Upsilon_a \\ \Upsilon_b \end{bmatrix} \in \mathbb{C} \quad (2)$$

Simulation 2: Alamouti-based coding

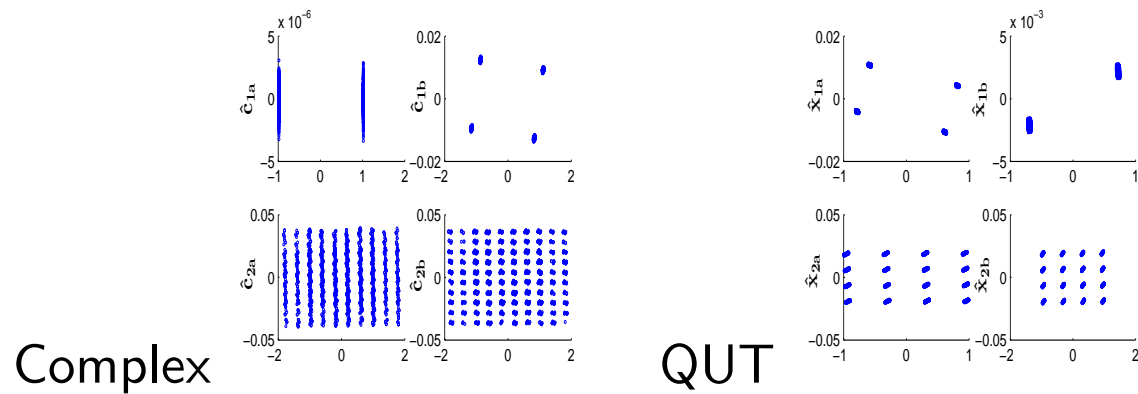
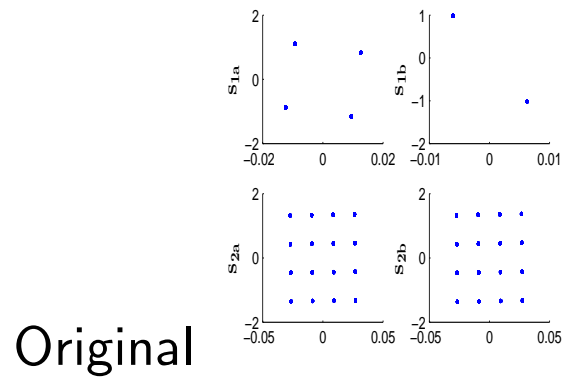
- Model for two users:

$$\begin{aligned} \begin{bmatrix} \mathbf{x}_{1a} \\ \mathbf{x}_{1b} \end{bmatrix} &= \begin{bmatrix} \mathbf{s}_{1a} & -\mathbf{s}_{1b}^* \\ \mathbf{s}_{1b} & \mathbf{s}_{1a}^* \end{bmatrix} \begin{bmatrix} \Upsilon_{11a} \\ \Upsilon_{11b} \end{bmatrix} + \begin{bmatrix} \mathbf{s}_{2a} & -\mathbf{s}_{2b}^* \\ \mathbf{s}_{2b} & \mathbf{s}_{2a}^* \end{bmatrix} \begin{bmatrix} \Upsilon_{12a} \\ \Upsilon_{12b} \end{bmatrix} \in \mathbb{C} \\ \begin{bmatrix} \mathbf{x}_{2a} \\ \mathbf{x}_{2b} \end{bmatrix} &= \begin{bmatrix} \mathbf{s}_{1a} & -\mathbf{s}_{1b}^* \\ \mathbf{s}_{1b} & \mathbf{s}_{1a}^* \end{bmatrix} \begin{bmatrix} \Upsilon_{21a} \\ \Upsilon_{21b} \end{bmatrix} + \begin{bmatrix} \mathbf{s}_{2a} & -\mathbf{s}_{2b}^* \\ \mathbf{s}_{2b} & \mathbf{s}_{2a}^* \end{bmatrix} \begin{bmatrix} \Upsilon_{22a} \\ \Upsilon_{22b} \end{bmatrix} \in \mathbb{C} \quad (3) \\ \begin{bmatrix} \mathbf{x}_1 \\ \mathbf{x}_2 \end{bmatrix} &= \begin{bmatrix} \Upsilon_{11} & \Upsilon_{12} \\ \Upsilon_{21} & \Upsilon_{22} \end{bmatrix} \begin{bmatrix} \mathbf{s}_1 \\ \mathbf{s}_2 \end{bmatrix} \in \mathbb{H} \quad \begin{cases} \mathbf{x}_i = \mathbf{x}_{i_a} + \mathbf{x}_{i_b} \\ \mathbf{s}_i = \mathbf{s}_{i_a} + \mathbf{s}_{i_b} \end{cases} \end{aligned}$$

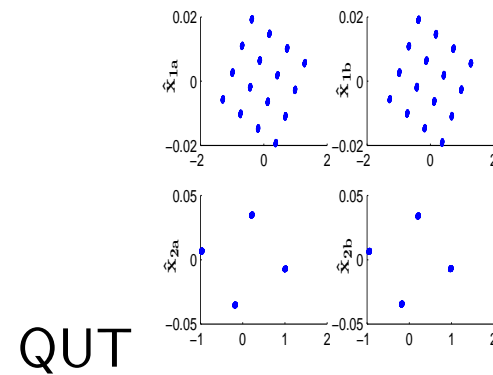
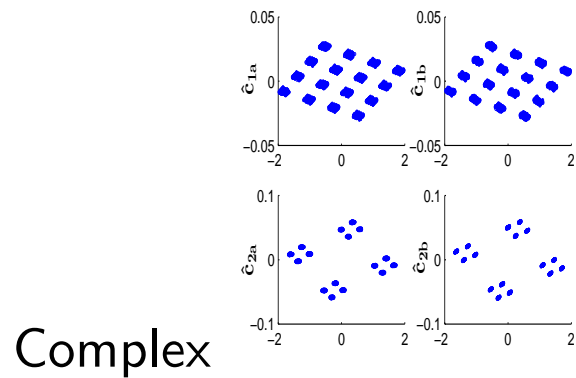
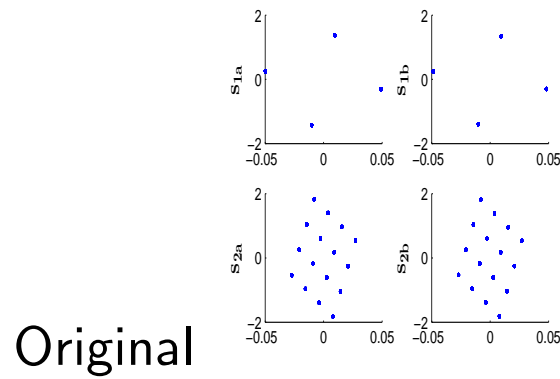
Quaternion-valued mixing matrix Υ and $s_{1a} = s_{1b}$



Complex-valued mixing matrix Υ and $s_{1a} \neq s_{1b}$



Complex-valued mixing matrix Υ and $s_{1a} = s_{1b}$



Conclusion

1. A set of matrix decompositions for the joint diagonalisation of quaternion covariance matrices which satisfy:
 - $\mathbf{AB} = \mathbf{BA}$
 - normality of \mathbf{AB}^η
 - $\mathbf{BA} = \mathbf{A}^\eta\mathbf{B}$
2. MATLAB code available at my website²
3. Open problem 1: How to diagonalise a quaternion symmetric matrix $\mathbf{A} = \mathbf{A}^T$?
4. Open problem 2: How to simultaneously diagonalise three and four quaternion matrices with different structures?

²MATLAB code available at http://www.surrey.ac.uk/cs/people/clive_cheong_took/index.htm

References

Thank you for your kind attention...

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1. Generate three ARMA sources $\mathbf{s}_d \in \mathbb{H}^{1 \times N}$, $d = \{1, 2, 3\}$
 2. To obtain uncorrelated \mathbb{C}^ν -proper sources, their real and imaginary parts were considered as real-valued signals and the following steps were applied:

1. $\mathbf{G} = [\mathbf{s}_{1r}; \mathbf{s}_{1i}; \mathbf{s}_{1j}; \mathbf{s}_{1\kappa}; \mathbf{s}_{2r}; \mathbf{s}_{2i}; \mathbf{s}_{2j}; \mathbf{s}_{2\kappa}; \mathbf{s}_{3r}; \mathbf{s}_{3i}; \mathbf{s}_{3j}; \mathbf{s}_{3\kappa}]$

2. $\mathbf{C}_G = \mathbf{G}\mathbf{G}^T$

3. $[\mathbf{U}, \mathbf{\Lambda}] = \text{svd}(\mathbf{C}_G)$

4. $\tilde{\mathbf{G}} = \mathbf{\Lambda}^{-\frac{1}{2}} \mathbf{U}^T \mathbf{G}$

5. Produce the desired covariance matrix $\mathbf{C}_D \in \mathbb{R}^{12 \times 12}$

$$\mathbf{C}_D = \left[\begin{array}{c|c|c} \mathbf{C}_s & \mathbf{0} & \mathbf{0} \\ \hline \mathbf{0} & \mathbf{C}_s & \mathbf{0} \\ \hline \mathbf{0} & \mathbf{0} & \mathbf{C}_s \end{array} \right], \quad \mathbf{C}_s = \begin{bmatrix} 1 & 0 & 0 & \rho_\nu \\ 0 & 1 & \rho_\nu & 0 \\ 0 & \rho_\nu & 1 & 0 \\ \rho_\nu & 0 & 0 & 1 \end{bmatrix}$$

in which the zero matrix $\mathbf{0}$ is used to provide uncorrelatedness, and $\mathbf{C}_s \in \mathbb{R}^{4 \times 4}$ is used to satisfy the \mathbb{C}^ν -properness with specific ν -circularity coefficient.

6. $[\mathbf{U}_D, \mathbf{\Lambda}_D] = \text{svd}(\mathbf{C}_D)$

7. $\mathbf{S} = \mathbf{U}_D \mathbf{\Lambda}_D^{\frac{1}{2}} \tilde{\mathbf{G}} = [\mathbf{s}_{1r}; \mathbf{s}_{1i}; \mathbf{s}_{1j}; \mathbf{s}_{1\kappa}; \mathbf{s}_{2r}; \mathbf{s}_{2i}; \mathbf{s}_{2j}; \mathbf{s}_{2\kappa}; \mathbf{s}_{3r}; \mathbf{s}_{3i}; \mathbf{s}_{3j}; \mathbf{s}_{3\kappa}]$

8. $\mathbf{s}_d = \mathbf{s}_{dr} + \nu \mathbf{s}_{di} + j \mathbf{s}_{dj} + \kappa \mathbf{s}_{d\kappa}$, $d = \{1, 2, 3\}$

3. Using the above sources, generate inputs \mathbf{x}_1 , \mathbf{x}_2 , and \mathbf{x}_3

Consider three univariate quaternion-valued sources as:

$$\begin{cases} \mathbf{x} = \mathbf{x}_r + \iota\mathbf{x}_\iota + j\mathbf{x}_j + \kappa\mathbf{x}_\kappa \\ \mathbf{y} = \mathbf{y}_r + \iota\mathbf{y}_\iota + j\mathbf{y}_j + \kappa\mathbf{y}_\kappa \\ \mathbf{z} = \mathbf{z}_r + \iota\mathbf{z}_\iota + j\mathbf{z}_j + \kappa\mathbf{z}_\kappa \end{cases} \in \mathbb{H}^{1 \times N} \quad (4)$$

To obtain the real-valued covariance matrix, consider the quadrivariate correspondence of each source ($\mathbf{x}_{\mathbb{R}}$, $\mathbf{y}_{\mathbb{R}}$, $\mathbf{z}_{\mathbb{R}}$) and generate the 12×12 covariance matrix in \mathbb{R} :

$$\mathbf{W} = \begin{bmatrix} \mathbf{x}_{\mathbb{R}} \\ \mathbf{y}_{\mathbb{R}} \\ \mathbf{z}_{\mathbb{R}} \end{bmatrix} = \begin{bmatrix} \mathbf{x}_r \\ \mathbf{x}_\iota \\ \mathbf{x}_j \\ \mathbf{x}_\kappa \\ \mathbf{y}_r \\ \mathbf{y}_\iota \\ \mathbf{y}_j \\ \mathbf{y}_\kappa \\ \mathbf{z}_r \\ \mathbf{z}_\iota \\ \mathbf{z}_j \\ \mathbf{z}_\kappa \end{bmatrix} \rightarrow \mathbf{C}_{\mathbf{w}} = E[\mathbf{W}\mathbf{W}^T] = E \left[\begin{array}{c|c|c} \mathbf{x}_{\mathbb{R}}\mathbf{x}_{\mathbb{R}}^T & \mathbf{x}_{\mathbb{R}}\mathbf{y}_{\mathbb{R}}^T & \mathbf{x}_{\mathbb{R}}\mathbf{z}_{\mathbb{R}}^T \\ \mathbf{y}_{\mathbb{R}}\mathbf{x}_{\mathbb{R}}^T & \mathbf{y}_{\mathbb{R}}\mathbf{y}_{\mathbb{R}}^T & \mathbf{y}_{\mathbb{R}}\mathbf{z}_{\mathbb{R}}^T \\ \mathbf{z}_{\mathbb{R}}\mathbf{x}_{\mathbb{R}}^T & \mathbf{z}_{\mathbb{R}}\mathbf{y}_{\mathbb{R}}^T & \mathbf{z}_{\mathbb{R}}\mathbf{z}_{\mathbb{R}}^T \end{array} \right] \quad (5)$$

$$E \begin{bmatrix}
x_r x_r^T & x_r x_i^T & x_r x_j^T & x_r x_\kappa^T & x_r y_r^T & x_r y_i^T & x_r y_j^T & x_r y_\kappa^T & x_r z_r^T & x_r z_i^T & x_r z_j^T & x_r z_\kappa^T \\
x_i x_r^T & x_i x_i^T & x_i x_j^T & x_i x_\kappa^T & x_i y_r^T & x_i y_i^T & x_i y_j^T & x_i y_\kappa^T & x_i z_r^T & x_i z_i^T & x_i z_j^T & x_i z_\kappa^T \\
x_j x_r^T & x_j x_i^T & x_j x_j^T & x_j x_\kappa^T & x_j y_r^T & x_j y_i^T & x_j y_j^T & x_j y_\kappa^T & x_j z_r^T & x_j z_i^T & x_j z_j^T & x_j z_\kappa^T \\
x_\kappa x_r^T & x_\kappa x_i^T & x_\kappa x_j^T & x_\kappa x_\kappa^T & x_\kappa y_r^T & x_\kappa y_i^T & x_\kappa y_j^T & x_\kappa y_\kappa^T & x_\kappa z_r^T & x_\kappa z_i^T & x_\kappa z_j^T & x_\kappa z_\kappa^T \\
\hline
y_r x_r^T & y_r x_i^T & y_r x_j^T & y_r x_\kappa^T & y_r y_r^T & y_r y_i^T & y_r y_j^T & y_r y_\kappa^T & y_r z_r^T & y_r z_i^T & y_r z_j^T & y_r z_\kappa^T \\
y_i x_r^T & y_i x_i^T & y_i x_j^T & y_i x_\kappa^T & y_i y_r^T & y_i y_i^T & y_i y_j^T & y_i y_\kappa^T & y_i z_r^T & y_i z_i^T & y_i z_j^T & y_i z_\kappa^T \\
y_j x_r^T & y_j x_i^T & y_j x_j^T & y_j x_\kappa^T & y_j y_r^T & y_j y_i^T & y_j y_j^T & y_j y_\kappa^T & y_j z_r^T & y_j z_i^T & y_j z_j^T & y_j z_\kappa^T \\
y_\kappa x_r^T & y_\kappa x_i^T & y_\kappa x_j^T & y_\kappa x_\kappa^T & y_\kappa y_r^T & y_\kappa y_i^T & y_\kappa y_j^T & y_\kappa y_\kappa^T & y_\kappa z_r^T & y_\kappa z_i^T & y_\kappa z_j^T & y_\kappa z_\kappa^T \\
\hline
z_r x_r^T & z_r x_i^T & z_r x_j^T & z_r x_\kappa^T & z_r y_r^T & z_r y_i^T & z_r y_j^T & z_r y_\kappa^T & z_r z_r^T & z_r z_i^T & z_r z_j^T & z_r z_\kappa^T \\
z_i x_r^T & z_i x_i^T & z_i x_j^T & z_i x_\kappa^T & z_i y_r^T & z_i y_i^T & z_i y_j^T & z_i y_\kappa^T & z_i z_r^T & z_i z_i^T & z_i z_j^T & z_i z_\kappa^T \\
z_j x_r^T & z_j x_i^T & z_j x_j^T & z_j x_\kappa^T & z_j y_r^T & z_j y_i^T & z_j y_j^T & z_j y_\kappa^T & z_j z_r^T & z_j z_i^T & z_j z_j^T & z_j z_\kappa^T \\
z_\kappa x_r^T & z_\kappa x_i^T & z_\kappa x_j^T & z_\kappa x_\kappa^T & z_\kappa y_r^T & z_\kappa y_i^T & z_\kappa y_j^T & z_\kappa y_\kappa^T & z_\kappa z_r^T & z_\kappa z_i^T & z_\kappa z_j^T & z_\kappa z_\kappa^T
\end{bmatrix} \tag{6}$$

If three uncorrelated \mathbb{C}^{ℓ} -proper sources are generated, the following observations on the structure of the covariance can be made:

- The off diagonal blocks in $\mathbf{C}_{\mathbf{w}}$ are zero, since sources are uncorrelated with regard to each other.
- The diagonal blocks have the following algebraic structure in order to satisfy the \mathbb{C}^{ℓ} -properness condition with specific values ρ .

$$\mathbf{C}_{\mathfrak{s}} = \begin{bmatrix} 1 & 0 & 0 & \rho \\ 0 & 1 & \rho & 0 \\ 0 & \rho & 1 & 0 \\ \rho & 0 & 0 & 1 \end{bmatrix} \quad (7)$$

Based on the matrix $\mathbf{C}_{\mathfrak{s}}$, the covariance matrices of each source are defined as:

$$\mathbf{C}_{\mathbf{q}} = 4\mathbf{I}, \quad \mathbf{C}_{\mathbf{q}^{\ell}} = \kappa(4\rho)\mathbf{I}, \quad \mathbf{C}_{\mathbf{q}^j} = \mathbf{C}_{\mathbf{q}^{\kappa}} = \mathbf{0}, \quad \mathbf{q} \in \{\mathbf{x}, \mathbf{y}, \mathbf{z}\} \quad (8)$$

The final structure of the covariance matrix is given by:

$$\mathbf{C}_{\mathbf{w}} = \begin{bmatrix} \mathbf{C}_{\mathfrak{s}} & \mathbf{0} & \mathbf{0} \\ \mathbf{0} & \mathbf{C}_{\mathfrak{s}} & \mathbf{0} \\ \mathbf{0} & \mathbf{0} & \mathbf{C}_{\mathfrak{s}} \end{bmatrix} \quad (9)$$

Table 1: The standard and η -covariance matrices of \mathbf{q} in terms of its real components.

| | $\mathbf{C}_{\mathbf{q}}$ | $\mathbf{C}_{\mathbf{q}^i}$ |
|------------------------------|---|---|
| $\mathfrak{R}[\cdot]$ | $E[\mathbf{q}_r \mathbf{q}_r^T] + E[\mathbf{q}_i \mathbf{q}_i^T] + E[\mathbf{q}_j \mathbf{q}_j^T] + E[\mathbf{q}_\kappa \mathbf{q}_\kappa^T]$ | $E[\mathbf{q}_r \mathbf{q}_r^T] + E[\mathbf{q}_i \mathbf{q}_i^T] - E[\mathbf{q}_j \mathbf{q}_j^T] - E[\mathbf{q}_\kappa \mathbf{q}_\kappa^T]$ |
| $\mathfrak{I}_i[\cdot]$ | $E[\mathbf{q}_i \mathbf{q}_r^T] - E[\mathbf{q}_r \mathbf{q}_i^T] + E[\mathbf{q}_\kappa \mathbf{q}_j^T] - E[\mathbf{q}_j \mathbf{q}_\kappa^T]$ | $E[\mathbf{q}_i \mathbf{q}_r^T] - E[\mathbf{q}_r \mathbf{q}_i^T] + E[\mathbf{q}_j \mathbf{q}_\kappa^T] - E[\mathbf{q}_\kappa \mathbf{q}_j^T]$ |
| $\mathfrak{I}_j[\cdot]$ | $E[\mathbf{q}_j \mathbf{q}_r^T] - E[\mathbf{q}_r \mathbf{q}_j^T] + E[\mathbf{q}_i \mathbf{q}_\kappa^T] - E[\mathbf{q}_\kappa \mathbf{q}_i^T]$ | $E[\mathbf{q}_r \mathbf{q}_j^T] + E[\mathbf{q}_j \mathbf{q}_r^T] - E[\mathbf{q}_\kappa \mathbf{q}_i^T] - E[\mathbf{q}_i \mathbf{q}_\kappa^T]$ |
| $\mathfrak{I}_\kappa[\cdot]$ | $E[\mathbf{q}_\kappa \mathbf{q}_r^T] - E[\mathbf{q}_r \mathbf{q}_\kappa^T] + E[\mathbf{q}_j \mathbf{q}_i^T] - E[\mathbf{q}_i \mathbf{q}_j^T]$ | $E[\mathbf{q}_\kappa \mathbf{q}_r^T] + E[\mathbf{q}_r \mathbf{q}_\kappa^T] + E[\mathbf{q}_i \mathbf{q}_j^T] + E[\mathbf{q}_j \mathbf{q}_i^T]$ |
| | $\mathbf{C}_{\mathbf{q}^j}$ | $\mathbf{C}_{\mathbf{q}^\kappa}$ |
| $\mathfrak{R}[\cdot]$ | $E[\mathbf{q}_r \mathbf{q}_r^T] - E[\mathbf{q}_i \mathbf{q}_i^T] + E[\mathbf{q}_j \mathbf{q}_j^T] - E[\mathbf{q}_\kappa \mathbf{q}_\kappa^T]$ | $E[\mathbf{q}_r \mathbf{q}_r^T] - E[\mathbf{q}_i \mathbf{q}_i^T] - E[\mathbf{q}_j \mathbf{q}_j^T] + E[\mathbf{q}_\kappa \mathbf{q}_\kappa^T]$ |
| $\mathfrak{I}_i[\cdot]$ | $E[\mathbf{q}_i \mathbf{q}_r^T] + E[\mathbf{q}_r \mathbf{q}_i^T] + E[\mathbf{q}_\kappa \mathbf{q}_j^T] + E[\mathbf{q}_j \mathbf{q}_\kappa^T]$ | $E[\mathbf{q}_i \mathbf{q}_r^T] + E[\mathbf{q}_r \mathbf{q}_i^T] - E[\mathbf{q}_j \mathbf{q}_\kappa^T] - E[\mathbf{q}_\kappa \mathbf{q}_j^T]$ |
| $\mathfrak{I}_j[\cdot]$ | $E[\mathbf{q}_j \mathbf{q}_r^T] - E[\mathbf{q}_r \mathbf{q}_j^T] - E[\mathbf{q}_\kappa \mathbf{q}_i^T] + E[\mathbf{q}_i \mathbf{q}_\kappa^T]$ | $E[\mathbf{q}_r \mathbf{q}_j^T] + E[\mathbf{q}_j \mathbf{q}_r^T] + E[\mathbf{q}_\kappa \mathbf{q}_i^T] + E[\mathbf{q}_i \mathbf{q}_\kappa^T]$ |
| $\mathfrak{I}_\kappa[\cdot]$ | $E[\mathbf{q}_\kappa \mathbf{q}_r^T] + E[\mathbf{q}_r \mathbf{q}_\kappa^T] - E[\mathbf{q}_i \mathbf{q}_j^T] - E[\mathbf{q}_j \mathbf{q}_i^T]$ | $E[\mathbf{q}_\kappa \mathbf{q}_r^T] - E[\mathbf{q}_r \mathbf{q}_\kappa^T] + E[\mathbf{q}_i \mathbf{q}_j^T] - E[\mathbf{q}_j \mathbf{q}_i^T]$ |