### Underdetermined DOA Estimation of Deterministic Signals Using High Order Statistics and Noncircularity

by

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To my parents

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## Abstract

Sensor arrays play important roles in signal transmission/reception, estimation, and tracking, and have been successfully applied to many engineering fields such as radar, sonar, wireless communications to name a few. Practically, sensor array systems usually suffer from nonideal factors such as signal coherency, spatially coloured noise, and a limited number of sensors. In this thesis, problems of direction of arrival (DOA) estimation in the presence of nonideal factors are addressed, and new algorithms to tackle these problems are developed that achieve improved performance with a limited number of sensors.

Under multipath propagation, independent and coherent signals coexist, resulting in rank deficiency of the cumulant matrix. To tackle this problem, two methods for DOA estimation of mixed independent and coherent signals using fourth-order cumulants (FOC) are proposed, and both algorithms can make efficient use of the array degrees of freedom (DOFs). The first algorithm implements the estimation via two-stage processing by separating the independent and coherent signals. In this method, new matrix reconstruction techniques for independent signal cumulants and rank restoration are developed, and the DOAs of both the independent and coherent signals can be estimated by polynomial rooting without performing a spectral grid search. Its superiority over existing methods is demonstrated by simulation results.

The second algorithm considers the case when a large number of coherent signals, greater than the number of sensors, exist due to the propagation channel. Here, we exploit temporal correlation in the signals to form an array output matrix with pseudo snapshots, spanning the same signal subspace as the one using real snapshots. By incorporating this property, new augmented cumulant matrices are constructed and the corresponding method for coherent group separation is derived. Compared with the existing method, the proposed one achieves better performance in terms of estimation accuracy and robustness of the spatial signature, especially for weak signals.

Apart from signal with circular statistics discussed above, we study the noncircularity embedded in modern wireless communication signals to further extend the effective aperture, enhance DOFs, and improve the estimation performance. A new FOC-based direction finding method which can extend the array aperture as well as maximise the DOFs is proposed. By combining noncircularity with high order cumulants and optimsing geometric arrangement of the virtual array arising accordingly, the resultant identifiability of DOA estimation can be up to twice larger compared with the using the same order cumulants for circular signals. Simulation results validate that the proposed method offers better performance in terms of identifiability as well as accuracy.

Last, we revisit the case when uncorrelated and coherent signals coexist and utilise the noncircularity of signals in this scenario. To the best of our knowledge, there are no publications addressing the class of DOA estimation problem, and a novel two-stage second order statistics (SOS) estimator is introduced accordingly to further increase the DOFs. In this method, a more robust approach is presented to identify the true DOA estimates from the pseudo ones, the estimates of noncircular phases are derived in closed-form, and a novel spatial smoothing technique based on the eigenvectors is developed to restore the rank deficiency. Additionally, new deterministic Cramér-Rao lower bounds (CRLBs) are derived for the considered mixture model of noncircular signals. The theoretical analysis justifies that the number of identifiable signals is larger than the current algorithms. Extensive simulation results show that the proposed method offers sufficient DOFs as well as improving the estimation accuracy of both the uncorrelated and coherent signals.

# **Statement of Originality**

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Date

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# Conventions

### Typesetting

This thesis is typeset using the LATEX2e software.

The fonts used in this thesis are Times New Roman and Sans Serif.

### Referencing

Referencing and citation style in this thesis are based on the Institute of Electrical and Electronics Engineers (IEEE) Transaction style [1].

For electronic references, the last accessed date is shown at the end of a reference.

### Units

The units used in this thesis are based on the International System of Units (SI units) [2].

### Spelling

The Australian English spelling is adopted in this thesis.

### **Publications**

### Journal

- Y. Wang, M. Trinkle, and B. W.-H. Ng, "DOA estimation under unknown mutual coupling and multipath with improved effective array aperture," *Sensors*, vol. 15, no. 12, pp. 30856–30869, Dec. 2015.
- [2] Y. Wang, M. Trinkle, and B. W.-H. Ng, "Two-stage DOA estimation of independent and coherent signals in spatially coloured noise," *Signal Process.*, vol. 128, pp. 350–359, Nov. 2016.
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### Conference

- Y. Wang and M. Trinkle, "DOA estimation for coherent signals with symmetric virtual array," in International Global Navigation Satellite Systems Symposium, IGNSS 2013, Outrigger Gold Coast, Australia, Jul. 2013.
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# **Abbreviations**

AIC	Akaike Information Criterion
АМ	Amplitude Modulation
АР	Alternating Projection
AR	Autoregressive
ARMA	Autoregressive Moving Average
ASK	Amplitude Shift Keying
BPSK	Binary Phase Shift Keying
СВ	Conventional Beamforming
ССІ	Co-Channel Interferences
CRLB	Cramér-Rao Lower Bound
CW	Continuous Wave
DML	Deterministic Maximum Likelihood
DOA	Direction Of Arrival
DOFs	Degrees Of Freedom
EM	Expectation-Maximization
ESPRIT	Estimation of Signal Parameter by Rotation Invariance Techniques
FBSS	Forward/Backward Spatial Smoothing
FOC	Fourth Order Cumulants
GDE	Gerschgorin Disk Estimator

#### Abbreviations

нос	High Order Cumulants
HOS	High Order Statistics
LL	Log-Likelihood
MDL	Minimum Description Length
ML	Maximum Likelihood
MLE	Maximum Likelihood Estimator
MRAs	Minimum Redundancy Arrays
MSE	Mean Squared Error
MUSIC	MUltiple Signal Classification
MVDR	Minimum Variance Distortionless Response
NSF	Noise Subspace Fitting
OQPSK	Offset Quadrature Phase Shift Keying
PAM	Pulse Amplitude Modulation
PDF	Probability Density Function
RMSE	Root Mean Squared Error
SAGE	
	Space Alternating Generalized EM
SF	Space Alternating Generalized EM Subspace Fitting
SF SML	
	Subspace Fitting
SML	Subspace Fitting Stochastic Maximum Likelihood
SML SNR	Subspace Fitting Stochastic Maximum Likelihood Signal-to-Noise Ratio

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- **SSF** Signal Subspace Fitting
- **SVD** Singular Value Decomposition
- **TAM** Toeplitz Approximation Method
- **ULA** Uniform Linear Array
- **VESPA** Virtual ESPRIT Algorithm
- **WSSF** Weighted Signal Subspace Fitting

### Notations

λ	The wavelength of corresponding frequency
θ	Azimuth angle
С	The speed of light
d	The spacing between adjacent sensors
$\mathbf{x} \in \mathbb{R}^M$	A vector with dimension <i>M</i> in real domain
$\mathbf{x} \in \mathbb{C}^M$	A vector with dimension <i>M</i> in complex domain
$(\ \cdot\ )^T$	The transpose operation
$(\cdot)^*$	The conjugate operation
$(\ \cdot\ )^H$	The conjugate transpose operation
$(\cdot)^{-1}$	The inverse operation
$( \cdot )^+$	The pseudo-inverse operation
[·]	The ceiling operation of a decimal number
[·]	The flooring operation of a decimal number
$E_t[\cdot]$	Statistical expectation implemented by $\lim_{L \to \infty} \frac{1}{L} \sum_{t=1}^{L} [\cdot]$
$\frac{\partial f(\mathbf{x})}{\partial \mathbf{x}}$	The first derivative of the function $f(\mathbf{x})$ with respect to $\mathbf{x}$
$\frac{\partial^2 f(\mathbf{x})}{\partial \mathbf{x}^2}$	The second derivative of the function $f(\mathbf{x})$ with respect to $\mathbf{x}$
$\operatorname{Re}\{\cdot\}$	Real part of a complex number
$Im\{\cdot\}$	Imaginary part of a complex number
$span\{\cdot\}$	Linear span
$\dim\{\cdot\}$	Dimension of linear space
$\mathcal{R}(\cdot)$	Range (or signal space) of a matrix
$\mathcal{N}(\cdot)$	Null space of a matrix
$\mathcal{V}^{\perp}$	Subspace orthogonal to a linear space $\mathcal{V}$
$\mathbf{P}_{\mathbf{A}}^{\perp}$	Orthogonal projection to the range of A
$E_{A B}$	Oblique projection with the range $\mathcal{R}(A)$ and null space $\mathcal{N}(B)$
$\max\{a, b\}$	The maximum value between <i>a</i> and <i>b</i>
$\min_{x} f(x)$	Minimise the function $f$ with respect to the variable $x$
$\ \cdot\ _2$	Euclidean ( $\ell_2$ ) norm

$\ \cdot\ _F$	Frobenius norm
$det\{\cdot\}$	Matrix determinant
0	Khatri-Rao product
$\otimes$	Kronecker product
$\odot$	Schur-Hadamard product, i.e., element-wise product
$\mathrm{tr}\{\cdot\}$	Matrix trace
$rank(\cdot)$	Matrix rank
diag $\{z_1, z_2\}$	Diagonal matrix with diagonal entries $z_1, z_2$
blkdiag{ $Z_1, Z_2$ }	Block diagonal matrix with diagonal entries $\mathbf{Z}_1, \mathbf{Z}_2$
$\mathbf{Z}(a:b,c:d)$	Submatrix by the entries from rows $a$ to $b$ and columns $c$ to $d$ of <b>Z</b>
$\mathbf{Z}(a,b)$	Entry in the $a$ -th row and $b$ -th column of $Z$
Ι	Identity matrix
J	Exchange matrix
R	Covariance matrix
$\mathbf{U}_{s}$	Signal subspace
$\mathbf{U}_n$	Noise subspace
a( heta)	Steering vector
$\mathbf{A}( heta)$	Array manifold

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