

Over-the-Horizon Radar Array Calibration

by

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Abstract

Modern over-the-horizon radars are currently being developed which have arrays than can be erected quickly, with minimal site preparation. Due to the rapid deployment of these arrays, antenna/sensor position errors may be present. Further since the antennas have a simple and cost-effective design, mutual coupling may be present. These imperfections, which can degrade radar performance, form the basis for the work conducted in this thesis.

The effect of these model errors on radar performance is first analysed. The degradation in signal-to-noise ratio, array gain, bearing estimation and array sidelobe levels, are determined. The major degradation is observed in the array sidelobe levels, which in turn results in the clutter-to-noise ratio (and hence target detectability) being worsened in the presence of non-stationary interferences. For these reasons, array calibration is required to improve the array sidelobe levels.

New array calibration algorithms are then developed to correct for sensor position errors and mutual coupling, using sources in the radar environment. These algorithms are analysed using simulations, and are found to perform well. The Cramer-Rao lower bound is derived, for the problem scenarios considered, and the algorithms are shown to achieve the bound. Further the Cramer-Rao lower bound is analysed, to obtain useful insight into the array calibration problem and identifiability.

Scattered echoes from meteor trails are shown to be excellent sources of opportunity for over-the-horizon radar array calibration. These echoes are found in general to : have planar wavefronts, be present in large numbers, be sufficiently strong, and be of adequate duration for sufficient snapshots to be obtained for array calibration. It also is shown that meteor head echoes are good sources of opportunity, and their properties along with that of other sources, are determined.

Finally, the receiving array of the Jindalee over-the-horizon radar (located in central Australia) is calibrated using echoes from meteor trails. The results obtained are compared with standard calibration (for this radar), which involves the use of special calibration sources. Array calibration using meteor trail echoes is found to perform as well as the standard array calibration, indicating that echoes from meteor trails can perform good array calibration.

Declaration

This work contains no material which has been accepted for the award of any other degree or diploma in any university or other tertiary institution and, to the best of my knowledge and belief, contains no material previously published or written by another person, except where due reference has been made in the text.

I give consent to this copy of my thesis, when deposited in the University Library, being available for loan and photocopying.

Signature

Date _____ April, 1998

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Glossary

Abbreviations

ASL	average sidelobe level	
C-TLS	constrained total least squares	
CRLB	Cramer-Rao lower bound	
DOAs	directions-of-arrival	
ESPRIT	estimation of signal parameters via rotational invariance techniques	
FIM	Fisher information matrix	
ISLB	integrated sidelobe level	
LS	least squares	
ML	maximum likelihood	
MUSIC	multiple signal classification	
PSL	peak sidelobe level	
SCV	sub-clutter visibility	
SLB	sidelobe	
SNR	signal-to-noise ratio	
STD	standard deviation	
TLS	total least squares	

Symbols

$\mathbf{x}^{T}, \mathbf{X}^{T}$	transpose
$\mathbf{x}^{H}, \mathbf{X}^{H}$	conjugate transpose
$\hat{x}, \hat{\mathbf{x}}, \hat{\mathbf{X}}$	estimates
$trace\{\mathbf{X}\}$	trace of matrix X
$\Re\{x\}$	real part of x
$\Im\left\{ x ight\}$	imaginary part of x
$conj(x), x^*$	conjugate of x
$diag\{\mathbf{x}\}, \mathbf{D}_{\mathbf{x}}$	diagonal matrix containing the elements of \mathbf{x} along its main diagonal
$ \mathbf{x} $	vector norm of x
$\ \mathbf{X}\ _F$	Frobenious norm of matrix \mathbf{X}
$\frac{\partial Q}{\partial x}$	partial derivative of Q with respect to x
\dot{x}_y	first derivative of x with respect to y
e^x , $\exp(x)$	exponential operator
$\delta(t_1,t_2)$	one when $t_1 = t_2$, and zero otherwise
$(\mathbf{X})^+$	generalised inverse of matrix \mathbf{X}
\odot	Hadamard product
j	complex operator
\mathbf{I}_M	$M \mathbf{x} M$ identity matrix
M	number of sensors / receivers
t	time index
Т	number of snapshots/samples
N	number of signals
N_C	number of clusters
N_T	total number of signals in all clusters
$N_{ ilde{Z}}$	number of columns in matrix $\tilde{\mathbf{Z}}$
$ heta,\phi$	bearing (azimuth), direction-of-arrival
Φ	set of bearings not in main beamwidth
(x_m, y_m)	actual position of mth sensor
(x^o_m,y^o_m)	nominal position of mth sensor
$(\Delta x_m, \Delta y_m)$	position error of m th sensor
λ	radar wavelength

SYMBOLS

f	radar frequency
w	radar angular frequency
υ	speed of light
$ au_{mn}$	time delay for n th signal to arrive at m th sensor
$\mathbf{a}(heta)$	array steering vector
$\mathbf{a}_o(heta)$	nominal steering vector
v	actual steering vector (including the effects of mutual coupling)
Α	matrix of array steering vectors
$\mathbf{D}(heta), \mathbf{D}_2(heta)$	diagonal matrices involved in the derivatives of $\mathbf{a}(\theta)$
$s_n(t)$	baseband waveform of the n th signal at time t
$s_I(t)$	baseband waveform of interference signal at time t
$n_m(t)$	receiver noise at time t
$z_m(t)$	output of m th sensor/receiver at time t
$\mathbf{s}(t)$	vector of baseband waveforms of signals, at time t
$\mathbf{n}(t)$	vector of noise outputs at time t
$\mathbf{n}_T(t)$	vector of total noise contribution (i.e. including interferers) at time t
$\mathbf{z}(t)$	vector of sensor outputs at time t
$\mathbf{z}_C(t)$	clutter signal at time t
$ar{\mathbf{z}}_m(t)$	output of consecutive receivers starting at m th receiver, at time t
$\tilde{\mathbf{Z}}$	matrix with data from all clusters
$\check{\mathbf{z}}(t)$	vector of sensor outputs for all sources, at time t
$\widetilde{z}_{ heta}(t)$	beamformer output for a beam steered in direction θ , at time t
$\mathbf{w}(heta)$	beamformer weight vector
$B_ heta(\phi)$	be ampattern for steer direction θ
$ASL(\theta)$	average sidelobe level for a beam steered in direction θ
p(heta)	beamformer power output in direction θ
$\sigma_S^2, \sigma_N^2, \sigma_I^2$	signal, noise, and interferer powers
σ_n^2	power of nth disjoint source
$s_n, \mathbf{s}, \mathbf{S}$	complex parameters
R	spatial covariance matrix
\mathbf{R}_n	spatial covariance matrix of nth disjoint source/cluster
Р	signal covariance matrix
\mathbf{P}_n	signal covariance matrix of n th disjoint cluster

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Q	noise covariance matrix
\mathbf{v}_m, \mathbf{V}	mth eigenvector and matrix of eigenvectors
λ_m	mth eigenvalue
E	signal subspace
U	noise subspace
$lpha_m,\psi_m$	gain and phase error, for m th element
Г	diagonal matrix of receiver gain/phase errors
σ_a	standard deviation of amplitude errors
σ_ψ	standard deviation of phase errors
\mathbf{S}_C	scattering matrix
С	coupling matrix
\mathbf{C}_{o}	nominal coupling matrix
\mathbf{Z}_{o}	array impedance matrix
Z_L	scalar load impedance
Q	cost function
н	Hessian matrix
r	gradient vector
Ψ	unknowns to be estimated
J	Fisher information matrix
$\mathbf{CRLB}(\mathbf{\Psi})$	Cramer-Rao lower bound for estimating Ψ

Boldface lower case variables are column vectors. Boldface upper case variables are matrices.

Publications

The list of publications relating to this thesis are :

- I. S. D. Solomon, Yu. I. Abramovich, D. A. Gray and S. J. Anderson, "OTH radar antenna array calibration analysis", Fourth International Symposium on Signal Processing and its Applications, August 1996, Gold Coast, Australia, pp. 471-474.
- I. S. D. Solomon, D. A. Gray, Yu. I. Abramovich and S. J. Anderson, "Estimating of array mutual coupling and sensor positions for over-the-horizon radar", Digital Signal Processing Applications Conference (TENCON), November 1996, Perth, Australia, pp. 846-851.
- I. S. D. Solomon, D. A. Gray, Yu. I. Abramovich and S. J. Anderson, "Sources for OTH radar array calibration", IEEE Antennas and Propagation Symposium, July 1997, Montreal, Canada, pp. 306-309.
- I. S. D. Solomon, D. A. Gray, Yu. I. Abramovich and S. J. Anderson, "OTH radar array calibration using disparate sources", IEE Radar Conference, October 1997, Edinburgh, Scotland, pp. 176-180.
- I. S. D. Solomon, D. A. Gray, Yu. I. Abramovich and S. J. Anderson, "Over-the-horizon radar array calibration using echoes from ionised meteor trails", IEE Proceedings Radar, Sonar and Navigation, June 1998, Vol. 145, No. 3, pp. 173-180 (to appear).
- I. S. D. Solomon, D. A. Gray, Yu. I. Abramovich and S. J. Anderson, "Array calibration of OTH radars using disparate sources", submitted to IEEE Transactions on Antennas and Propagation.
- I. S. D. Solomon, Yu. I. Abramovich, D. A. Gray and S. J. Anderson, "Performance of OTH radar array calibration", IEEE International Conference on Acoustics, Speech, and

Signal Processing (ICASSP), May 1998, Seattle, Washington, U.S.A., pp 2025-2028.

- I. S. D. Solomon, Yu. I. Abramovich, D. A. Gray and S. J. Anderson, "Mutual coupling estimation", International Radar Symposium, September 1998, Munich, Germany (to appear).
- I. S. D. Solomon, D. A. Gray, Yu. I. Abramovich and S. J. Anderson, "Meteor trail echoes for OTH radar array calibration", *in preparation for* Radio Science.