# Multiple Signal Detection using the ESPRIT Algorithm

Alfred Tsz Yin Lok, Zekeriya Aliyazicioglu, H. K. Hwang

Abstract— An array antenna system with innovative signal processing to estimate the direction of arrival (DOA) for multiple targets is investigated in this paper. The signal processing technique in this study is the ESPRIT (estimation signal parameter via a rotational invariant technique). The DOA angles are derived indirectly from the generalized eigenvalues of the auto-correlation and cross-correlation matrices. Two different methods to estimate the multi-signal's DOA are: (1) Choice of a proper set of generalized eigenvalues from the two sub-arrays and (2) Choice of an appropriate proper set of generalized eigenvectors from the two sub-arrays. The estimated autocorrelation and cross correlation matrices in this study is based on combination of temporal averaging and spatial smoothing method. Extensive computer simulations are used to demonstrate the performance of the processing algorithms.

*Index Terms*—DOA estimation, array antenna, advanced signal processing.

#### I. INTRODUCTION

Accurately estimating the direction of arrival (DOA) has many important applications in communication and radar systems. Using the conventional fixed antenna, the resolution of DOA is limited by the antenna mainlobe beamwidth. Using the array antenna and advanced signal processing techniques, the DOA estimation variance can be greatly reduced.

Two important classes of signal processing techniques are the model based approach and the eigenanalysis method[1]. The model based method assumes that the received data is modeled as the output of a linear shift invariant system. The DOA information may be obtained indirectly from the estimated model parameters. Several eigen-analysis methods such as multiple signal classification (MUSIC)[2], polynomial root intersection for multidimensional estimation (PRIME)[3,4] have been investigated by several authors. This paper is a study of multi-signal's DOA estimation using the ESPRIT[5,6,7] algorithm. DOA estimation of a single

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Z. Aliyazicioglu is with the Electrical and Computer Engineering Department, California State Polytechnic University, Pomona, CA 91768 USA, phone: 909-869-3667; fax: 909-869-4687; (e-mail: zaliyazici@ csupomona.edu).

H. K. Hwang is with the Electrical and Computer Engineering Department, California State Polytechnic University, Pomona, CA 91768 USA phone: 909-869-2539; fax: 909-869-4687; (e-mail: hkhwangr@csupomona.edu). target using ESPRIT method is fairly straightforward. A signal's DOA can be derived from the maximum generalized eigenvalue[6] of two independent equations. If there are multiple targets, determining the signal DOA depends on the choice of an appropriate pair of generalized eigenvalues from two independent equations. If the appropriate pair of generalized eigenvalues is not chosen, it results in a false DOA. Rather than choosing appropriate pairs based on the generalized eigenvalues, improved results can be obtained by choosing the appropriate pair based on the generalized vectors. A performance comparison is presented in this paper.

The array antenna used in this simulation study consists of 19 elements in a honeycomb configuration. In this paper DOA performance is discussed as a function of signal to noise ratio (SNR), and as an effect of spatial smoothing[8].

#### II. ESPRIT ALGORITHM

The array antenna considered in this paper has 19 antenna elements in a honeycomb configuration as shown in Figure 1. Array elements are uniformly placed on an x-y plane. The inter-element spacing d equals half of the signal wavelength.



Figure 1 Two Dimensional Arrays with 19 Elements

Assume the n<sup>th</sup> narrowband signal is impinging on the array from an elevation angle  $\theta_n$  and azimuth angle  $\phi_n$  as shown in Figure 2. Using the n<sup>th</sup> signal received by the center element  $s_{nc}(t)$  as the reference, the n<sup>th</sup> signal received by the i<sup>th</sup> element  $s_{ni}(t)$  is Proceedings of the World Congress on Engineering 2010 Vol II WCE 2010, June 30 - July 2, 2010, London, U.K.



Figure 2 Coordinate of array system and signal direction

$$s_{ni}(t) = s_{nc}(t) e^{j\beta_{ni}}$$
<sup>(1)</sup>

where the electrical angle of the i<sup>th</sup> element  $\beta_{ni}$  is

$$\beta_{ni} = -\frac{2\pi}{\lambda} \sin\theta (x_i \cos\varphi_n + y_i \sin\varphi_n)$$
(2)

where  $(x_i, y_i)$  are the coordinates of the i<sup>th</sup> element.

Equation (2) shows that the signal DOA angles  $(\theta_n, \phi_n)$  are related to the electrical angle  $\beta_{ni}$ . The ESPRIT algorithm derives the DOA angles from the phase factor  $\beta$ . Determining two angles  $(\theta_n, \phi_n)$  requires two different phase factors. Two independent phase factors can be derived from two independent position shifts. A brief description of ESPRIT using a 19 element array is as follows:

An appropriate subset is chosen from the original array and this subset is shifted in two different directions. The subset in this study consists of element (1, 2, 4, 5). The data vectors of the subset and shifted subsets (2, 3, 5, 6) and (4, 5, 8, 8) are related by the appropriate electrical angles. The signal's elevation and azimuth angles can be derived from the two electrical angles corresponding to the two different shifts. The subset and two different shifts used in this study are shown in Figure 3.



Figure 3. Subset and Two Linearly Shifted Subsets

Now , assume there are two independent narrowband signals impinging on the antenna from DOA  $(\theta_1, \phi_1)$  and  $(\theta_2, \phi_2)$ , respectively. The waveform **y** received by the subset

ISBN: 978-988-18210-7-2 ISSN: 2078-0958 (Print); ISSN: 2078-0966 (Online) consists of elements (1, 2, 4, 5) and can be expressed as:

$$\mathbf{y}(n) = s_{1,10}(n)\mathbf{s}_1 + s_{2,10}(n)\mathbf{s}_2 + \mathbf{w}_y(n)$$
(3)

where  $\mathbf{w}_{y}(n)$  is the random white noise and  $\mathbf{w}_{y}(n) = [\mathbf{w}_{1}(n), \mathbf{w}_{2}(n), \mathbf{w}_{4}(n), \mathbf{w}_{5}(n)]^{T}$ ,  $\mathbf{s}_{1,10}(n)$  is the signal received by the center element due to signal 1,  $\mathbf{s}_{2,10}(n)$  is the signal received by center element due to signal 2,  $\mathbf{s}_{1} = [e^{j\beta_{1,1}}, e^{j\beta_{1,2}}, e^{j\beta_{1,4}}, e^{j\beta_{1,5}}]^{T}$ ,  $\mathbf{s}_{2} = [e^{j\beta_{2,1}}, e^{j\beta_{2,2}}, e^{j\beta_{2,4}}, e^{j\beta_{2,5}}]^{T}$ . The correlation matrix  $\mathbf{R}_{yy}$  of this subset is

$$\mathbf{R}_{yy} = \mathrm{E}[\mathbf{y}\mathbf{y}^{\mathrm{H}}] = \begin{bmatrix} \mathbf{s}_{1} & \mathbf{s}_{2} \begin{bmatrix} P_{1} & 0\\ 0 & P_{2} \end{bmatrix} \begin{bmatrix} \mathbf{s}_{1}^{\mathrm{H}}\\ \mathbf{s}_{2}^{\mathrm{H}} \end{bmatrix} + \sigma_{\mathrm{w}}^{2} \mathbf{I} \qquad (4)$$

where  $P_1$  and  $P_2$  are the power of signal 1 and 2, respectively and  $\sigma_w^2$  is the noise variance.

Shifting this subset horizontally to the right forms a new subset consisting of elements (2, 3, 5, 6). The received waveform of this new subset z(n) is

$$\mathbf{z}(n) = s_{1,10}(n)\mathbf{s}_3 + s_{2,10}(n)\mathbf{s}_4 + \mathbf{w}_z(n)$$
(5)

where  $\mathbf{w}_{z}(n) = [w_{2}(n), w_{3}(n), w_{5}(n), w_{6}(n)]^{T}$  is the noise vector,  $\mathbf{s}_{3} = e^{j\beta_{1,11}} \mathbf{s}_{1}, \mathbf{s}_{4} = e^{j\beta_{2,11}} \mathbf{s}_{2}.$ 

The cross correlation matrix  $\mathbf{R}_{yz}$  is

$$\mathbf{R}_{yz} = \mathbf{E}[\mathbf{y}\mathbf{z}^{H}] = \begin{bmatrix} \mathbf{s}_{1} & \mathbf{s}_{2} \begin{bmatrix} P_{1} & 0\\ 0 & P_{2} \end{bmatrix} \begin{bmatrix} e^{-j\beta_{1,1}} \mathbf{s}_{1}^{H}\\ e^{-j\beta_{2,1}} \mathbf{s}_{2}^{H} \end{bmatrix} + \sigma_{w}^{2} \mathbf{Q} \quad (6)$$
  
where  $\mathbf{Q} = \begin{bmatrix} 0 & 0 & 0 & 0\\ 1 & 0 & 0 & 0\\ 0 & 0 & 0 & 0\\ 0 & 0 & 1 & 0 \end{bmatrix}$ 

Arranging the eigenvalues of matrix  $\mathbf{R}_{yy} \lambda_1$ ,  $\lambda_2$ ,  $\lambda_3$ ,  $\lambda_4$  in descending order, the noise variance  $\sigma_w^2$  can be estimated by the following equation.

$$\sigma_{\rm w}^2 = (\lambda_3 + \lambda_4)/2 \tag{7}$$

Define matrices  $C_{yy}$  and  $C_{yz}$  as

$$\mathbf{C}_{yy} = \mathbf{R}_{yy} - \boldsymbol{\sigma}_{w}^{2} \mathbf{I} = \begin{bmatrix} \mathbf{s}_{1} & \mathbf{s}_{2} \end{bmatrix} \begin{bmatrix} \mathbf{P}_{1} & \mathbf{0} \\ \mathbf{0} & \mathbf{P}_{2} \end{bmatrix} \begin{bmatrix} \mathbf{s}_{1}^{H} \\ \mathbf{s}_{2}^{H} \end{bmatrix}$$
(8)

$$\mathbf{C}_{yz} = \mathbf{R}_{yz} - \boldsymbol{\sigma}_{w}^{2} \mathbf{Q} = \begin{bmatrix} \mathbf{s}_{1} & \mathbf{s}_{2} \begin{bmatrix} P_{1} & 0\\ 0 & P_{2} \end{bmatrix} \begin{bmatrix} e^{-j\beta_{1,1}} \mathbf{s}_{1}^{H}\\ e^{-j\beta_{2,1}} \mathbf{s}_{2}^{H} \end{bmatrix}$$
(9)

Then  $\mathbf{C}_{yy} - \lambda \mathbf{C}_{yz} = \begin{bmatrix} \mathbf{s}_1 & \mathbf{s}_2 \begin{bmatrix} P_1 & 0 \\ 0 & P_2 \end{bmatrix} \begin{bmatrix} \left(1 - \lambda e^{-j\beta_{1,1}}\right) \mathbf{s}_1^H \\ \left(1 - \lambda e^{-j\beta_{2,1}}\right) \mathbf{s}_2^H \end{bmatrix}$ , thus  $\lambda =$ 

 $e^{j\beta_{1,11}}$  and  $e^{j\beta_{2,11}}$  are two of the roots of det( $C_{yy} - \lambda C_{yz}$ ). Two phase factors  $\beta_{1,11}$  and  $\beta_{2,11}$  can be obtained by finding the

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two roots of det( $C_{yy} - \lambda C_{yz}$ ) closest to the unit circle.

Forming subset (4, 5, 8, 9), the second independent phase factor can be obtained. The received data vector of this subset **v** is:

$$\mathbf{v}(n) = e^{j\beta_{1,14}} \mathbf{s}_{1,10}(n) \mathbf{s}_1 + e^{j\beta_{2,14}} \mathbf{s}_{2,10}(n) \mathbf{s}_2 + \mathbf{w}_v(n)$$
(10)

where  $\mathbf{w}_{v}(n) = [w_4(n), w_5(n), w_8(n), w_9(n)]^T$  is the noise vector.

The cross correlation matrix  $\mathbf{R}_{yy}$  is

$$\mathbf{R}_{yv} = \mathbf{E}[\mathbf{yv}^{\mathrm{H}}] = \begin{bmatrix} \mathbf{s}_{1} & \mathbf{s}_{2} \begin{bmatrix} P_{1} & 0\\ 0 & P_{2} \end{bmatrix} \begin{bmatrix} e^{-j\beta_{1,4}} \mathbf{s}_{1}^{\mathrm{H}} \\ e^{-j\beta_{2,14}} \mathbf{s}_{2}^{\mathrm{H}} \end{bmatrix}^{+} \sigma_{w}^{2} \mathbf{Q}_{1}$$
(11)  
where  $\mathbf{Q}_{1} = \begin{bmatrix} 0 & 0 & 0 & 0\\ 0 & 0 & 0 & 0\\ 1 & 0 & 0 & 0\\ 0 & 1 & 0 & 0 \end{bmatrix}$ 

Define matrices  $C_{yy}$  and  $C_{yy}$  as:

$$\mathbf{C}_{yy} = \mathbf{R}_{yy} - \boldsymbol{\sigma}_{w}^{2} \mathbf{I} = \boldsymbol{\sigma}_{s}^{2} \mathbf{s} \mathbf{s}^{H}$$
(12)

$$\mathbf{C}_{yv} = \mathbf{R}_{yv} - \boldsymbol{\sigma}_{w}^{2} \mathbf{Q}_{1} = \boldsymbol{\sigma}_{s}^{2} \mathbf{e}^{j\beta_{s}} \mathbf{ss}^{H}$$
(13)

Then  $\mathbf{C}_{yy} - \lambda \mathbf{C}_{yv} = \begin{bmatrix} \mathbf{s}_1 & \mathbf{s}_2 \begin{bmatrix} P_1 & 0\\ 0 & P_2 \end{bmatrix} \begin{bmatrix} (1 - \lambda e^{-j\beta_{1,14}}) \mathbf{s}_1^H\\ (1 - \lambda e^{-j\beta_{2,14}}) \mathbf{s}_2^H \end{bmatrix}$ , thus  $\lambda = \mathbf{s}_1 = \mathbf{s}_2 \begin{bmatrix} \mathbf{s}_1 & \mathbf{s}_2 \end{bmatrix}$ 

 $e^{j\beta_{1,14}}$  and  $e^{j\beta_{2,14}}$  are two of the roots of det( $C_{yy} - \lambda C_{yv}$ ). Two phase factors  $e^{j\beta_{1,14}}$  and  $e^{j\beta_{2,14}}$  can be obtained by finding the two roots of det( $C_{yy} - \lambda C_{yv}$ ) closest to the unit circle.

From 
$$\beta_{n,11} = \pi \sin \theta_n \cos \phi_n$$
, (14)

$$\beta_{n,14} = -\frac{\pi}{2}\sin\theta_n(\cos\phi_n + \sqrt{3}\sin\phi_n)$$
(15)

The electrical angles  $\beta_{n,11}$ ,  $\beta_{n,14}$  for n = 1, 2 are obtained from the appropriate pair of roots closest to unit circle from det( $\mathbf{C}_{yy} - \lambda \mathbf{C}_{yz}$ ) and det( $\mathbf{C}_{yy} - \lambda \mathbf{C}_{yv}$ ). Once we determine the electrical angles  $\beta_{n,11}$ ,  $\beta_{n,14}$ , the signal DOA can be obtained by solving ( $\theta_n, \phi_n$ ) from Equations (14) and (15).

#### **III. MATRICES ESTIMATION**

Section 2 shows that the DOA angles are derived from the auto-correlation and cross-correlation matrices. DOA performance depends on the accurate estimation of matrices  $\mathbf{R}_{yy}$ ,  $\mathbf{R}_{yz}$ ,  $\mathbf{R}_{yy}$ . Elements of matrices are estimated from the received data y(n) = s(n) + w(n) where s(n) and w(n) are the signal and white noise of the received data. Three matrix estimation methods, (1) temporal averaging, (2) spatial smoothing, (3) temporal averaging and spatial smoothing, are described in this section [8].

# A. Temporal Averaging Method

This method estimates the matrix element r<sub>ij</sub> by averaging

ISBN: 978-988-18210-7-2 ISSN: 2078-0958 (Print); ISSN: 2078-0966 (Online) the products of data received by the i<sup>th</sup> and j<sup>th</sup> elements over N snapshots according to the following equation:

$$r_{ij} = \frac{1}{N} \sum_{n=1}^{N} y_i(n) y_j^*(n)$$
(16)

#### B. Spatial Smoothing Method

Since the number of elements in the array is larger than the size of the subset, instead of discarding the data from elements outside of the subset, those data can be used to improve the estimation of  $r_{ij}$ . Any pair of elements that has a similar geometrical relationship has the same spatial correlation. For example, elements of the correlation matrix of the square array for subset (1, 2, 4, 5) are computed from the following equations:

$$r_{11}(n) = r_{22}(n) = r_{44}(n) = r_{55}(n) = \frac{1}{19} \sum_{i=1}^{19} y_i(n) y_i^*(n)$$
(17)

There are 14 pairs of elements having the same correlation as  $r_{12}(n)$ , thus  $r_{12}(n)$  can be computed from the following equation.

$$r_{12}(n) = \frac{1}{14} \Big[ y_1(n) y_2^*(n) + y_2(n) y_3^*(n) + y_4(n) y_5^*(n) + \dots + y_{18}(n) y_{19}^*(n) \Big]$$
(18)

Other elements of the correlation matrix can be computed in a similar manner.

# C. Temporal Averaging and Spatial Smoothing Method

This method combines spatial smoothing and temporal averaging. After estimating the matrix elements  $r_{ij}(n)$  from spatial smoothing, an estimated  $r_{ij}$  is obtained by further averaging over N snapshots according to the following equation.

$$r_{ij} = \frac{1}{N} \sum_{n=1}^{N} r_{ij}(n)$$
(19)

## IV. SIMULATION RESULTS

Assume that two tone signals are impinging on the array from (10°, 50°) and (30°, 120°), where the first angle is the elevation angle and second angle is the azimuth angle. The signal to noise ratios of the first and second signals are 20 dB and 10 dB respectively. Since there are two signals, accurate DOA estimation relied on properly pairing the generalized eigenvalues of ( $C_{yy}$ ,  $C_{yz}$ ), and ( $C_{yy}$ ,  $C_{yv}$ ). Since the second signal power is 10 dB lower than the first one, we pair the generalized eigenvalues closest to unit circle to define the DOA of stronger signal and the next pair of generalized eigenvalues closest to the unit circle to define the DOA of the weaker signal. Using a 4 element subset of the 19 element antenna, Figure 4 shows the estimated DOA histogram plot based on 1000 independent simulations. The estimated matrices are based on temporal averaging over 32 snapshots. Proceedings of the World Congress on Engineering 2010 Vol II WCE 2010, June 30 - July 2, 2010, London, U.K.



Figure 4. Histogram Based on Temporal Averaging over 32 Snapshots

Figure 4 shows that the estimated DOA is quite noisy. This is due to the fact the estimated matrices are based on temporal averaging over 32 snapshots. Increasing the number of snapshots should improve the DOA estimation. However, when operating the system in real time the number of snapshots is usually limited. Improved DOA performance can be achieved by matrices estimation using a combination of temporal averaging and spatial smoothing. Figure 5 shows the estimated DOA histogram by matrices estimation using temporal averaging and spatial smoothing.



Figure 5. Histogram Based on Temporal Averaging over 32 Snapshots and Spatial Smoothing

Figure 5 shows that the two clear peaks are very close to the signal DOA. The stronger peak is close to the stronger signal DOA at  $(10^{\circ}, 50^{\circ})$ . Figures 4 and 5 show that enhanced matrices estimation by combination of temporal averaging and spatial smoothing improves the performance of ESPRIT algorithm.

If the two signals have the same power, estimating the signal's DOA using an inappropriate pair of eigenvalues may create a false DOA. Figure 6 shows the histogram of two signals with the identical SNR = 20 dB.



Figure 6. Histogram Based on Temporal Averaging over 32 Snapshots and Spatial Smoothing

Since the two signals have the same amount of power, the probability of estimating signal DOA using an incorrect pair of generalized eigenvalues is fairly high. Thus Figure 6 shows that there are two false peaks due to having chosen the wrong pair of generalized eigenvalues.

Rather than define the correct pair of eigenvalues based on their relative distance from the unit circle, we can use the generalized eigenvectors to define the appropriate pair. Suppose the generalized eigenvalues of  $(\mathbf{C}_{yy}, \mathbf{C}_{yz})$  are  $\lambda_1, \lambda_2$ and their associated eigenvectors are  $\mathbf{q}_1$  and  $\mathbf{q}_2$ , the generalized eigenvalues of  $(\mathbf{C}_{yy}, \mathbf{C}_{yz})$  are  $\lambda_a$ ,  $\lambda_b$  and their associated eigenvectors are  $\mathbf{q}_a$  and  $\mathbf{q}_b$ . If the Euclidean distance between  $\mathbf{q}_1$  and  $\mathbf{q}_a$  is less than the Euclidean distance between  $\mathbf{q}_1$  and  $\mathbf{q}_b$ , then we pair  $(\lambda_1, \lambda_a)$  and  $(\lambda_2, \lambda_b)$  to compute the signal DOA. Otherwise we pair  $(\lambda_1, \lambda_b)$  and  $(\lambda_2, \lambda_a)$  to compute the signal DOA. Based on this method, the estimated DOA histogram is shown in Figure 7.



Figure 7. Histogram Based on Temporal Averaging over 32 Snapshots and Spatial Smoothing

Comparing Figures 6 and 7, the DOA estimation based on eigenvectors is seen to provide an improvement. The probability of detecting the false target is greatly reduced. Signals impinging the array from different directions have quite different eigenvectors. Thus choosing the proper pair of eigenvalues based on minimum Euclidean distance of the corresponding eigenvectors provides a much better DOA estimation. If there are more than two signals, estimating signal's DOA requires a larger subset. We will continue pursuit study of more than two signals. Proceedings of the World Congress on Engineering 2010 Vol II WCE 2010, June 30 - July 2, 2010, London, U.K.

# V. CONCLUSION

The conclusions based on the results of this simulation study are summarized as follows:

- 1. The ESPRIT method estimates signal DOA by finding the roots of two independent equations closest to the unit circle. This method does not require using a scan vector to scan over all possible directions like the MUSIC (Multiple Signal Classification) algorithm.
- 2. Performance of DOA estimation can be enhanced by improved matrices estimation. One method to improve matrices estimation is to estimate matrices by combination of temporal averaging and spatial smoothing.
- 3. When there is a large difference between the powers of two signals, choosing the first pair of eigenvalues as the one closest to unit circle and the second pair of eigenvalues as the second closest to unit circle provides a reasonable DOA estimation.
- 4. When the powers of two signals are identical, pairing the eigenvalues based only on their distance to the unit circle may create false peaks. An improved DOA estimation can be obtained by choosing the pair of eigenvalues whose associated eigenvectors have minimum Euclidean distance.
- 5. Array element position may deviate from the ideal position. Position deviation will degrade DOA performance. Sensitivity analysis due to imprecise element position will be carried out in a future study.

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