## Effect of varying Number of Spacing between Antenna Elements, Snapshots, SNR on AP ML, AP-SSF and ESPRIT Algorithm

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*Abstract:* - The direction of arrival estimation is the main key problem in array signal processing. In this paper, the alternating projection maximum Likelihood (AP-ML), Alternating projection sub space framework (AP-SSF) and ESPRIT algorithm are studied. The simulation is performed in MATLAB for single and multiple sources. The effect of the varying number of spacing between antenna elements, number of snapshots and SNR are studied. The performance comparison shows that ESPRIT algorithm performs better as compared to the AP-ML and AP-SSF.

Key-Words: - AP-ML, AP-SSF, Direction of Arrival, ESPRIT, Snapshots, SNR.

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### **1** Introduction

The main aim of the array signal processing is to process the incoming signal from the different directions and increase the signal strength by removing the noise and interference signal by collecting the desired signal parameter information. The ML methods have elite yet computationally costly. The subspace techniques are performed well and having computationally effective variations. The old style strategies are basic yet offer humble or lackluster showing and furthermore require an enormous number of calculations. In this the three algorithms are studied and the performance comparison of the three is studied in details.

# 2. Alternating Projection Maximum Likelihood

The AP-ML is a iterative based techniques for reducing the maximization problem from a multidimensional single dimensional to problem. This may be achieved bv maximization according to a single parameter keeping other parameters fixed. The maximum likelihood provides an optimum solution at an array of sensors for the direction finding of multiple signals. The computational complexity is high in maximum likelihood [2], [4].

By assuming an array of sensors (p) having different locations and q be the uncorrelated narrowband signals that are impinging on the array from the directions  $Q_1, Q_2, Q_3, \dots, Q_q$ . the received signal at the antenna array is given by:

$$x(t) = \sum_{m=1}^{q} a(Q_m) S_m(t) + n(t)$$
(1)

Where, x(t) is a  $p \times 1$  received signal vector, n(t) is a  $p \times 1$  white noise vector with zero mean and covariance matrix of  $\sigma^2 I \cdot \sigma^2$  and Ibe the unknown noise variance and identity matrix.  $S_m(t)$  be the signal emitted by the  $m^{th}$ source and received at a reference point.  $a(\theta_m)$ be the  $p \times 1$  steering vector corresponding to the direction  $\theta_m$ . The ML estimate is to solves the 1-D problem. The estimated angle is given by the following equation:

$$\theta_{i}^{\hat{k}+1} = \max_{\theta_{i}} \frac{a(\theta_{i})_{A(\theta_{i})k}^{H} Ra(\theta_{i})_{A\theta_{i}^{k}}}{a(\theta_{i})_{A\theta_{i}^{k}}^{H} a(\theta_{i})_{A\theta_{i}^{k}}}$$
(2)

where,  $a(\theta_i)_{A(\theta_i^k)}$  is a  $p \times 1$  vector. Given by  $a(\theta_i)_{A(\theta_i^k)} \equiv (I - PA_{\theta_i^k})a(\theta_i)$  (3)

Where,  $PA_{\theta_i^k}$  is a projection operator on to the

subspace spanned by the columns of  $A(\hat{\theta}_i)$  and H denotes the Hermitian Matrix.

# **3.** Alternating Projection SSF (Subspace Fitting Framework)

This alternating projection is used to estimate the different parameters in subspace fitting framework and gives the relationship between different algorithms. This type of framework is generally used to designed numerical algorithms and used in obtaining new methods. The extension of the ESPRIT algorithm is designed from the SSF method. The subspace fitting approach was firstly presented by [2] and it was formalized by [3] which is written as,

$$\begin{bmatrix} \hat{A}, \hat{T} \end{bmatrix} = \arg\min_{A,T} \|M - A(\eta)T\|_F^2$$
(4)

In this equation, *M* is a  $m \times q$  matrix that may be obtained from the given data. The  $m \times p$ matrix *A* is calculated from  $\eta$  and *T* is the  $p \times q$  matrix. The parameter estimation vector

 $\eta$  is the argument of  $\hat{A}$ . The matrix M and dimension of it can be chosen in the different ways to give different estimation. The fitting problem is separated with the help of A and T

From study, [5], putting A = A \* M in to the above equation, we may get the new equation,

$$\hat{A} = \arg\max_{A} T_r \left\{ P_A M M^* \right\}$$
(5)

Where,  $P_A = A(A * A)^{-1}A *$  is the projection matrix which is used to projects on the column space of A. Different algorithms and methods are explore like Deterministic Maximum Likelihood (DML), Beamforming, MUSIC, Multi dimensional MUSIC(MD-MUSIC), ESPRIT are proposed based on subspace fitting framework.

Alternating projection is used to transform the multivariate non-linear maximization problem in to a sequence of 1-d maximization.

The study [6] introduced a new algorithm by

maximizing the likelihood function given below:

$$L = -KdLog\sigma^{2} - \frac{1}{\sigma^{2}} \sum_{l=0}^{k-1} |x_{i}(l) - A(\theta)S_{i}|^{2}$$
 (6)

This method is known as alternating projection method. The flow chart and basic step follows in this method are given below. In this, firstly one dimensional projection is find out that is further used to maximize L.

### **4. ESPRIT (Estimation of Signal Parameters Via Rotational Invariance Techniques)**

The study, [1] proposed a new algorithm named ESPRIT for DOA estimation. Array doublets are formed by N/2 pairs which further form a displacement vector. The starting two elements of the doublet are separated and grouped to make two N/2 sub arrays. The vectors x and y are the data vectors corresponding to each of the sub arrays. The output of the sub arrays z and y can be expressed as:

$$x_{k}[n] = \sum_{i=0}^{r-1} s_{i}[n]a_{k}(\theta_{i}) + v_{k}^{(x)}[n], \qquad (7)$$

$$y_{k}[n] = \sum_{i=1}^{r-1} s_{i}[n] e^{j2\pi\delta\sin\theta_{k}} a_{k}(\theta_{i}) + v_{k}^{(y)}[n], \qquad (8)$$

Where similar notation has been used and  $\delta$  is displacement magnitude in wavelengths. The estimated angle by ESPRIT algorithm relative to the displacement vector. The sub arrays, x and y, output is given in matrix form is:

$$x_n = As_n + v_n^x \tag{9}$$

$$y_n = A\phi s_n + v_n^y \tag{10}$$

The matrix  $\phi$  is a diagonal  $r \times r$ , matrix having diagonal elements are  $\{\exp(j2\pi\delta\sin\theta_0), \exp(j2\pi\delta\sin\theta_1), ..., \exp(j2\pi\delta\sin\theta_{r-1})\}$ . The phase delay may be represented by the complex exponentials between the r signals and doublet pair. The data vectors may be concatenated from sub arrays to make a single 2N-2 data vector, like,

$$z_n = \begin{bmatrix} x_n \\ y_n \end{bmatrix} = A_b S_n \tag{11}$$

$$A_{b} = \begin{bmatrix} A \\ A\phi \end{bmatrix}, V_{n} = \begin{bmatrix} v_{n}^{(x)} \\ v_{n}^{(y)} \end{bmatrix}$$
(12)

The columns of  $A_b$  occupy the signal subspace of the new array. Let V<sub>s</sub> be the column matrix depending upon the signal subspace as Z<sub>n</sub>,  $A_b$ and V<sub>s</sub> are related with  $r \times r$  transformation T is written as:

$$V_s = A_b T, \tag{13}$$

and can portioned as follows:

$$V_{s} = \begin{bmatrix} E_{x} \\ E_{y} \end{bmatrix} = \begin{bmatrix} AT \\ A\phi T \end{bmatrix}$$
(14)

From this step, the range of  $E_x$ ,  $E_y$  and A will be equal as  $E_x$ ,  $E_y$  have the same range, the rank r matrix  $E_{xy}$  as follows:

$$E_{xy} = \begin{bmatrix} E_x & E_y \end{bmatrix}$$
(15)

To find  $r \times 2r$  rank r matrix having null space of  $E_{xy}$  to form matrix F, and is written as:

$$\begin{bmatrix} E_x & E_y \end{bmatrix} F = E_x F_x + E_y F_y = ATF_x + A\phi TF$$
(16) y

Assume  $\psi$  is:

$$\psi = -F_x [F_y]^{-1} \tag{17}$$

Reshuffling the above equations gives:

$$E_x \psi = E_y \tag{18}$$

Now by substituting we get the results:

$$AT\psi = A\phi T \Longrightarrow AT\psi T^{-1} = A\phi \Longrightarrow T\psi T^{-1}$$
(19) =  $\phi$ 

The given equations means that the Eigen values of  $\psi$  is same as diagonal elements of . Once the Eigen values,  $\lambda$ , of  $\phi$  have been calculated, the angle of arrival is calculated as:

$$\lambda_k = e^{j2\pi\delta\sin\theta_k} \tag{20}$$

$$\theta_k = \arcsin\left(\frac{\arg(\lambda_k)}{2\pi\delta}\right) \tag{21}$$

The flow chart of of AP-SSF algorithm is presented in Fig. 1. Similarly, the flow chart of ESPRIT Algorithm is presented in Fig. 2.



Figure 1: Flow chart of AP-SSF algorithm



### **Figure 2: ESPRIT Algorithm Flow Chart**

If A be the full rank matrix, then the Eigen values of the matrix  $\psi$  are the diagonal elements of and the Eigen vectors of  $\psi$  are the columns of the T. in practical, the signal subspace is not known exactly, the only estimate is from sample covariance matrix  $R_{xx}$  or from a sub space tracking algorithm. There for,  $E_x\psi = E_y$ , will not be exactly satisfied and we will have to resort to a least square solution to computes  $\psi$ . The least square process assumes that the columns in  $E_x$  are known

exactly whereas the data in  $E_y$  is noisy. If the assumptions is made that  $E_x$  and  $E_y$  are equally noisy, the total least square criteria is used to solve, which gives better results.

The above algorithms are studied in detailed and the comparison is given here for one, two, three, four and five signals.

Table 1: Estimated Signal Directions at receiving antenna array when number of elements = 10, number of samples = 100 and SNR = 20db

Algori	ithm	Signal Directions		
		(40°)		
Alternating ML	Projection	40.0120		
Alternating SSF	Projection	16.0040		
ESPRIT		40.0177		

Table 1 gives the signal estimation for single source that is coming from the direction 40° and the three algorithms the signal directions. The signal estimated from the AP-SSF gives wrong signal estimation while as comparison to the AL-ML, the ESPRIT algorithm gives better accuracy.

Table 2: Estimated Signal Directions at receiving antenna array when number of elements = 10, number of samples = 100 and SNR = 20db

Algorithm	Signal Directions	Signal Directions
	(-10°)	$(-20^{\circ})$
Alternating Projection ML	-9.8600	-19.9680
Alternating Projection SSF	-9.9480	-19.912
ESPRIT	-10.0312	-19.9641

Table 2 gives the description of the estimated signal when two signals from the directions  $-10^{\circ}$  and  $-20^{\circ}$  are coming at the receiver side and the number of antenna elements are take as 10, number of samples are 100 and SNR is 20dB.

**Table 3: Estimated Signal Directions at receiving** antenna array when number of elements=10, number of samples=100 and SNR=20db

Algorithm	Signal Direction s	Signal Direction s $(-20^{\circ})$	Signal Direction s $(-30^\circ)$			
	(-10)	(-20)	(-30)			
Alternatin g Projection ML	-3.3820	-26.0600	-30.0680			
Alternatin g Projection SSF	-7.0800	-21.4840	-30.4840			
ESPRIT	-10.0677	-20.4922	-30.1516			

Table 3 gives the description of the estimated signal when three signals from the directions  $-10^\circ$ ,  $-20^\circ$  and  $-30^\circ$  are coming at the receiver side and the number of antenna elements are take as 10, number of samples are 100 and SNR is 20dB.

**Table 4: Estimated Signal Directions at receiving** antenna array when number of elements = 10, number of samples = 100 and SNR = 20db

Algorithm	orithm Signal Directions		Signal Directions	
	(40°)	(50°)	(70°)	
Alternating Projection ML	40.0120	50.0080	70.0040	
Alternating Projection SSF	16.0040	66.0040	70.0040	
ESPRIT	40.0177	50.0052	69.9985	

Table 4 gives the Discription of the estimated signal when three signals from the directions  $40^{\circ}$ ,  $50^{\circ}$  and  $70^{\circ}$  are coming at side and the number the receiver of antenna elements are take as 10, number of samples are 100 and SNR is 20dB.

### **Table 5: Estimated Signal Directions at receiving** antenna array when number of elements = 10, number of samples = 64 and SNR = 20db

Algor ithm	Signal Directi ons (60°)	Signal Directio ns (80°)	Signa l Direc tions (90°	Signal Directions (100°)
Alter natin g	54.012 0	75.9960	92.99 20	99.9920

87.00

89.98

50

00

100.0080

100.0023

g Proje ction ML Alter

natin

ESP

RIT

g Proje ction SSF

60.004

60.013

1

0

Table 5, the simulations In are performed for the four signals when the number of array elements at the receiving antenna is considered 10, the samples taken at the receiving antenna are 64 and the signal to noise ratio at is 20dB.

76.0000

80.0166

**Table 6: Estimated Signal Directions at receiving** antenna array when number of elements=10, number of samples=100 and SNR=20db

Algori thm	Signal Directio ns (60°)	Signal Directi ons (80°)	Signal Directio ns (90°)	Signal Direction s (100°)
Altern ating Projec tion ML	54.0040	77.980 0	94.0200	99.9920
Altern ating Projec tion SSF	60.0080	77.992 0	89.0080	99.9880
ESPR IT	60.0067	79.989 1	90.0006	99.9832

In Table 6, the simulations are performed for the four signals when the number of array elements at the receiving antenna is considered 10, the samples taken at the receiving antenna are 100 and the signal to noise ratio at is 20dB. When we increase the number of snapshots, the accuracy of the algorithms is increased and the incoming signal direction is measured more accurate.

Table 7: Estimated Signal Directions at receiving antenna array when number of elements=10, number of samples=100 and SNR=20db

Algori thm	Signa l Direc tions (40°)	Signa l Direc tions (50°)	Signa l Direc tions (70°)	Signa l Direc tions (90°)	Signa l Direc tions (120°)
Altern ating Projec tion ML	36.02 40	42.01 20	77.36 00	104.9 920	120.2 400
Altern ating Projec tion SSF	39.99 20	47.99 60	64.02 00	85.98 80	119.9 960
ESPRI T	40.02 24	49.99 76	70.00 88	89.99 77	120.0 021

In Table 7, the simulations are performed for the five signals when the number of array elements at the receiving antenna is considered samples 10. the taken at the receiving antenna are 100 and the signal to noise ratio at is 20dB.

Table 8: Estimated Signal Directions at receiving antenna array when number of elements=100, number of samples=100 and SNR=20db

Algorith m	Signal Directi ons	Signal Directi ons	Signal Directi ons	Signal Directi ons
	(60°)	(80°)	(90°)	(100°)
Alternat ing Projecti	54.0000	80.0000	100.000 0	100.016 0

on ML				
Alternat ing Projecti on SSF	54.0000	54.0000	64.0000	100.000 0
ESPRIT	60.0009	80.0007	89.9998	99.9988

In Table 8, the simulations are performed for the four signals when the number of array elements at the receiving antenna is considered 100, the samples taken at the receiving antenna are 100 and the signal to noise ratio at is 20dB.

Table 9: Estimated Signal Directions at receiving antenna array when number of elements=5, number of samples=100 and SNR=20db

Algorith m	Algorith Signal m Directi ons		Signal Directi ons	Signal Directi ons	
	(60°)	(80°)	(90°)	(100°)	
Alternat ing Projecti on ML	23.9640	33.9960	95.5400	99.9640	
Alternat ing Projecti on SSF	60.0320	73.9520	90.6720	99.8840	
ESPRIT	60.0550	79.7319	90.0373	100.034 1	

In Table 9, the simulations are performed for the four signals when the number of array elements at the receiving antenna is considered 5, the samples taken at the receiving antenna are 100 and the signal to noise ratio at is 20dB.

# 5. Effect of increasing the number of elements, number of samples and SNR at antenna array

We studied the detailed comparison of the three algorithms when we increase the number of antenna elements at the receiver array, number of samples taken at the receiver side for signal estimation and by increasing the signal to noise ratio at the receiver side. The signals taken are two here form  $45^{\circ}$  and  $60^{\circ}$  directions respectively. The estimation through ESPRIT algorithm gives good results in all conditions when we increase the antenna elements, spacing between the antenna elements and SNR level.

#### Table 10: Estimated Signal Directions at receiving antenna array when numbers of elements are increasing keeping other parameters fixed.

Alg orit hm	Number of Element s=5 Number of Samples =100 SNP-20		lgNumberNumberNumbercitofofofmElementElementEls=5s=10s=NumberNumberofofofofSamplesSamplesSamples=100=1SNR=20SNR=20SNR=20		Num of Elen s=20 Num of Sam =100	Number of Element s=20 Number of Samples =100		Number of Element s=100 Number of Samples =100 SNR-20	
	dB Sig nal Dir ect ions	Sig nal Dir ect ions	dB Sig nal Dir ect ions	Sig nal Dir ect ions	dB Sig nal Dir ect ions	Sig nal Dir ect ions	dB Sig nal Dir ect ions	Sig nal Dir ect ions	
	45°	60°	45°	60°	45°	60°	45°	60°	
Alt ern ati ng Pro ject ion ML	45. 06 00	60. 03 60	45. 01 20	59. 99 60	45. 00 00	60. 00 00	45. 00 00	60. 00 00	
Alt ern ati ng Pro ject ion SS F	45. 05 20	60. 03 20	45. 01 20	59. 99 60	45. 00 00	60. 00 00	45. 00 00	60. 00 00	
ES PR IT	45. 06 26	60. 00 99	45. 01 77	59. 99 68	45. 00 00	60. 00 00	45. 00 00	60. 00 00	

In Table 10, the simulations are performed for the two signals incoming from the direction  $45^{\circ}$  and  $60^{\circ}$  when the number of array elements at the receiving antenna is increasing from 5, 10, 20, 100, the samples taken at the receiving antenna are 100 and the signal to noise ratio at is 20dB. From the above table it is clear that when we increase the number of samples the accuracy of the estimation is increased and the ESPRIT algorithm gives better results.

Table 11: Estimated Signal Directions at receiving antenna array when number of samples is increasing keeping other parameters

	-			fixed.	0	-		
Alg orit hm	Number of Element s=10		Number of Element s=10		Number of Element s=10		Number of Element s=10	
	Number of Samples =1		Number of Samples =10		Number of Samples =20		Number of Samples =100	
	SNR dB	=20	SNR dB	=20	SNR dB	=20	SNR dB	=20
	Sig nal Dir ect ion							
	S	S	S	S	S	S	S	S
	( 45° )	( 60° )	( 45° )	( 60° )	( 45° )	( 60° )	( 45° )	( 60° )
Alt ern ati ng Pro ject ion ML	45. 04 80	59. 92 40	44. 98 40	60. 01 60	45. 00 00	60. 00 80	45. 01 20	59. 99 60
Alt ern ati ng Pro ject ion SS	46. 22 80	65. 16 80	44. 97 60	60. 01 60	45. 00 00	60. 01 20	45. 01 20	59. 99 60
F								

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ES	49.	99.	44.	60.	44.	60.	45.	60.
PR	37	24	95	02	99	02	00	00
IT	30	65	12	40	57	01	00	00

In Table 11, the simulations performed are presented for the two signals incoming from the direction  $45^{\circ}$  and  $60^{\circ}$  when the number of samples taken at the receiver side is increasing from 1, 10, 20, 100, the number of antenna elements are fixed as 10 and the signal to noise ratio at is 20dB. From the above table it is clear that when we increase the number of antenna elements the accuracy of the estimation is increased.

 Table 12: Estimated Signal Directions at receiving

 antenna array when SNR are increasing keeping other

 parameters fixed.

Alg orit hm	Number of Elements =10 Number of Samples= 100 SNR=1d		Number of Elements =10 Number of Samples= 100 SNR=10d		Number of Elements =10 Number of Samples= 100 SNR=20d		Number of Elements =10 Number of Samples= 100 SNR=100	
	B Sig nal Dir ecti ons (	Sig nal Dir ecti ons (	B Sig nal Dir ecti ons (	Sig nal Dir ecti ons (	B Sig nal Dir ecti ons (	Sig nal Dir ecti ons (	dB Sig nal Dir ecti ons (	Sig nal Dir ecti ons (
	45°	60°	45°	60°	45°	60°	45°	60°
Alte rna ting Pro ject ion ML	45. 172 0	59. 976 0	45. 044 0	59. 992 0	45. 012 0	59. 996 0	45. 000 0	60. 000 0
Alte rna ting Pro ject ion SSF	45. 168 0	59. 972 0	45. 048 0	59. 984 0	45. 012 0	59. 996 0	45. 000 0	60. 000 0
ES PRI T	45. 124 6	59. 974 8	45. 052 6	59. 994 0	45. 000 0	60. 000 0	45. 000 0	60. 000 0

In Table 12, the simulations are performed for the two signals incoming from the direction  $45^{\circ}$  and  $60^{\circ}$  when the SNR at the receiver side is increasing from 1, 10, 20, 100, the number of antenna elements are fixed as 10 and the samples taken are 100.

When the SNR level is increased the estimation accuracy is decreased.

**Conclusion:** The simulation is performed in the MATLAB environment and the results obtained at different conditions. The performance comparison shows that ESPRIT algorithm performs better as compared to the AP-ML and AP-SSF.

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### **Conflicts of Interest**

The author(s) declare no potential conflicts of interest concerning the research, authorship, or publication of this article.

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