Synthesis of Elliptical Antenna Array using Hybrid SSWOA Algorithm

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Abstract - In terms of research, the elliptical antenna arrays (EAA) synthesis is relatively novel. As it does not have to be circular in construction, this novel synthesis can maneuver the primary beam in the right direction, making it easier to realize. The amplitude and angular location of the ellipse, as well as the eccentricity of the ellipse, are all taken into account in the optimization process. The proposed hybrid algorithm is the SSWOA (Salp Swarm Whale Optimization Algorithm), which combines the Salp Swarm Optimization Algorithm (SSA) with the Whale Optimization Algorithm (WOA). The SSA algorithm serves as a guide, while the WOA algorithm serves as a helper in this method. We discover that optimization has a faster convergence time and high convergence accuracy when considering the benefits of SSA and WOA and applying them to the synthesis of antenna array layouts. If Griewank, Rosenbrock, Sphere, and Rastrigin test functions are used, it's worth noting that the hybrid method outperforms both WOA and SSA.

Index Terms – elliptical antenna array (EAA), Griewank, pattern synthesis, speed of convergence, sphere, rastrigin, Rosenbrock.

I. INTRODUCTION

The single antenna structure of antenna arrays gives it a great advantage in its wide usage, especiallyin defence applications, for example, sonar, radar, communication, and navigation systems. Higher levels of gain and directivity can be provided by these antenna arrays. Generally, the antenna arrays are categorized based on geometrical differences. Concentric circular, planar, linear, and circular, besides elliptical arrays, can be cited as specimens of these clusters. Nevertheless, in their research area, elliptical antenna arrays are comparatively novel [1-6]. Various means are adapted to synthesize the elliptical antenna arrays. The majority of this work is focused on linear and circular antenna arrays. Nevertheless, it is a widely known fact that elliptical antenna arrays possess radiation properties just like that of circular antenna arrays. The entire space is occupied by the elliptical antenna arrays because of their radiation pattern; furthermore, the primary beam of the array can be changed to the desired direction. Linear antenna, rectangular arrays, and elliptical antenna arrays should not be coupled together as the latter is less sensitive. This is because they lack the edge elements [7-8]. The elliptical antenna array (EAA) synthesis is the combination

of the position only, the amplitude only, or amplitude position techniques. Of these, the most used technique is the position only technique. In these three distinct algorithms namely the SADE, i.e., self-adaptive differential evolution, the BBO, i.e. Biogeography-Based Optimization, in addition to Firefly algorithms (FA),[4] are used to identify the optimal position of elements on the ellipse circumference which has the capacity of bringing down the maximum side lobe level (SLL) with fixed Beam Width between First Nulls (BWFN). The amplitude technique is explored by making use of the non-uniform current distribution, where the element position is stationary and cannot be changed. Further, by comparing BSA with BBO, FA, genetic algorithm (GA), and differential evolution(DE) the performance of thealgorithm is evaluated.

In this paper, EAA radiation synthesis is designed using a hybrid algorithm of SSWOA. The SSA, i.e., Salp Swarm Optimization combined with WOA, i.e., Whale Optimization Algorithm is a newly suggested algorithm. In this algorithm, the SSA algorithm acts as a guide while the WOA algorithm serves the role of an aid. With the aim of preventing a premature convergence and exploring wholly the problem of hyperspace, the leader updating the neighbour's best point is measured. Likewise, after the salps point updating, by employing WOA to improve the optimum searchability, the well-known point hasevolved. A combination of these algorithms enhances the diversity of the offspring, as well as maintaining the search-balance.

This paper is organized as follows: The presentation of the radiation synthesis model employing a hybrid algorithm of SSWOA for EAA is dealt with in section II. In section III the comparisons and numerical experiments on the EAA pattern synthesis are illustrated. Section IV offers the concluding section of this paper.

II. METHODOLOGYPROPOSED

This section proposes the hybrid algorithm of SSWOA for radiation pattern synthesis of the EAA. The algorithm of SSWOA being proposed is a blend of WOA as well as SSA. In this, the SSA algorithm effectually guides the evolution, while the WOA algorithm serves as the assistant's role. With a view to averting an untimely convergence and with the view to fully exploring the problem in hyperspace, the leader updating the best position is wellthoughtout. A unique function is used to govern the algorithm proposed as a continuation of the above. An elaborative procedure of the algorithm being proposed is clearly explained in the following subsection.

A. The SSWOA algorithm

The SSWOA method is described and used in this part to create patterns for elliptical antenna arrays. It is

thought of as a global optimization problem,

min
$$f(X) = f(X_1, X_2, X_3, \dots, X_d);$$

for $X_i^{\min} \le X_i \le X_i^{\max},$ (1)

where i = 1, 2, 3, ..., d the number of optimized variables denoted as dX_i^{max} is the higher bound and X_i^{min} is the lower bound of X_i respectively.

As suggested in 2017 by Mirjalili [9], SSA is a new type of algorithm for resolving single-objective as well as multi-issues. The main motivation for the SSA is the swarming behaviour of salps when voyaging or foraging in the oceans. Finally, this SSA produces accurate and the best possible or even near-optimal arrangements in the optimization process. Particularly in this investigation, to give the optimal result competently, the investigative behaviour of SSA is increased by WOA [10– 11]. The WOA is a natural and meta-heuristic optimization algorithm, which is deduced from characteristics of humpback whales. The following steps delineate the complete process of the hybrid algorithm suggested.

Stage 1: Initiating

In the initial stage of the SSA, the places of the salps are arbitrarily instated in the dimensional space which is communicated as,

$$X_i = \text{rand} * (X_i^{max} - X_i^{min}) + L_i \text{ for } i = 1, 2, 3.....d.$$
(2)

Stage 2: Appraisal

In this step, the suitability of each explore agent is appraised. The subsequent equation provides the vital objective function,

$$fitness = \min\{f(X)\}.$$
 (3)

Stage 3: The procedure of search employing

Layer 1: Explorative Stage (searching for quarry)

In the Explorative Stage, the location of a search agent gets modernized.

$$B = \left| \overrightarrow{c} \cdot X_{rand}(i) - X(i) \right|, \tag{4}$$

$$X(i+1) = X_{rand}(i) - \overrightarrow{a} \cdot B.$$
(5)

The probing extract in equations (4) and (5) and by the shrinking system the encompassing quarry in equations (6) and (7) are employed if $\vec{a} \ge 1$.

Layer 2: Encompassing quarry

To discover an optimal solution proficiently, the probing performance of the SSA is boosted by the WOA, by the subsequent equation.

$$B = \left| \overrightarrow{c} \cdot X^* - X \right|, \tag{6}$$

$$X(i+1) = X^* - \overrightarrow{d} \cdot B, \tag{7}$$

where, \overrightarrow{c} and \overrightarrow{a} are considered as vector coefficients, and I as current iteration, the finest value of location

 $X_i = rand^*(X_i^{\max} - X_i^{\min}) + L_i$ Evaluate the fitness using $fitness = \min\{f(X)\}$ Searching process using WOA search for circling uvdating Laver Layer 2 Layer 3 Optimal solution

Random initialization

Fig. 1. Steps of proposed SSWOA algorithm.

vector up to now is symbolized like X^* and X denotes the location vector.

The vector coefficients \overrightarrow{c} and \overrightarrow{d} are expressed as,

$$\overrightarrow{c} = 2R, \tag{8}$$

$$d = 2A \times (R - A).$$
(9)
is juncture, the random numeral in the middle

At th of [0, 1] is specified like R and the changeable A is linearly diminishing as of [2-0].

Layer 3: Location apprising

The updating spiral location is represented as,

$$X(i+1) = \begin{cases} X_{rand}(i) - \overrightarrow{a} \cdot B; & if \ P < 0.5\\ B \cdot e^{hl} \cos(2\pi l) + X_{rand}(i); & if \ P \ge 0.5 \end{cases},$$
(10)

where the constant for portraying the spiral shape has been designated as h, the random numeral between [0, 1]is designated as P and 1 is [-1, 1].

Step 4: Termination procedure

When the procedure is accomplished, the projected technique chooses the optimum result or as well comes flip side to step 2. Figure 1 displays the steps of the projected hybrid algorithm of SSWOA. Thus, the grouping of SSA and WOA algorithms improves the variety of the progeny. Likewise, the desired search ability of the algorithm is heightened.

B. Elliptical antenna array (EAA) pattern synthesis array factor formulation

In this subsection, the geometry of EAA and the corresponding AF are explored, leading to array optimization in the following subsections. The EAA is a kind of antenna array in which the elements of the antenna are placed at the boundary of an ellipse. Figure 2 illustrates an instance of N-element EAA lying in the x-y plane and has its midpoint positioned at the origin [12].



Fig. 2. N-element EAA.

Array factor for the EAA portrayed in Fig. 2 is given by:

$$AF(\theta, \emptyset) = \sum_{(n=1)}^{N} I_n \exp(j[k\sin(\theta)](a\cos(\emptyset_n)\cos(\emptyset) + b\sin(\emptyset_n)\sin(\emptyset) + \alpha_n]).$$
(11)

Where:

a is the semi-major axis of the ellipse b is the semi-minor axis of the ellipse

 I_n is excitation amplitude of nth element

k= $2\pi/\lambda$ (wave number)

 θ is Elevation angle

 $Ø_n$ is angular position of the nth element,

$$\emptyset_n = \frac{2\pi(n-1)}{N}$$

Ø is azimuth angle

 α_n is excitation phase of the nth element

$$\alpha_n = k\sin(\theta 0)(a\cos(\emptyset n)\cos(\emptyset 0 + b\sin(\emptyset n)\sin[(\emptyset 0)]).$$
(12)

The eccentricity (e) of the ellipse is,

$$e = \sqrt{1 - \frac{b^2}{a^2}}.$$
 (13)

If the e lies between 0 and 1, when e is '0' then e becomes a circle, when e is '1', then it becomes a line.

Here, the radiation pattern synthesis of the array is discussed in detail to show the suggested hybrid algorithm. The array pattern is obtained through the superposition rule, where the radiation pattern of all active elements is extracted and stored. Several sub-goals put together are the primary objectives of the antenna array design to insert various objectives into one fitness function. This fitness function may be defined as,

Fitness function
$$= F_{f1} + F_{f2} + F_{f3} + F_{f4} + F_{ff} + F_{f6}$$
. (14)

At this point,

$$F_{f1} = |\phi_0 - \phi_d|,$$
 (15)

$$F_{f2} = |\theta_0 - \theta_d|, \tag{16}$$

$$F_{f3} = \begin{cases} |g - g_d| ; if g \le g_d \\ 0 ; if g > g_d \end{cases},$$
(17)

$$F_{f4} = \begin{cases} \left| MSL_L - SL_{L_d} \right| ; if MSL_L \ge SL_{L_d} \\ 0 ; if MSL_L < SL_{L_d} \end{cases}, \quad (18)$$

$$F_{f5} = \begin{cases} |NUL_L - NUL_{L_d}| ; if NUL_L \le NUL_{L_d} \\ 0 ; if NUL_L > NUL_{L_d} \end{cases}$$
(19)

$$F_{f6} = \begin{cases} |HP_{BW} - HP_{BWd}| ; if HP_{BW} \ge HP_{BWd} \\ 0 ; if HP_{BW} < HP_{BWd} \end{cases} (20)$$

The subsequent section displays the efficacy of the algorithm proposed with some numerical experimentation.

III. NUMERICAL RESULTS AND DISCUSSION

In this part, the adequacy of the projected calculation is checked to employ certain test capacities. At this point, we utilized four test capacities and these are utilized as central capacities for advancement techniques like the SSA besides the WOA. The capacities are clarified as follows,

$$Sphere_{(F_{1}(x))} = \sum_{i=1}^{N} X_{i}^{2};$$

for $|X_{i}| \le 100,$ (21)

Rosenbrock_{(F3(x))} =
$$\sum_{i=1}^{N-1} \left[100(X_{i+1} - X_i^2)^2 + (1 - X_i)^2 \right];$$

$$for |X_i| \le 30, \tag{22}$$

$$Rastrigin_{(F_4(x))} = \sum_{i=1}^{N} \left[X_i^2 - 10 \cos(2\pi X_i) + 10 \right];$$

for $|X_i| \le 10,$ (23)

$$Griewank_{(F_2(x))} = \left\{ \frac{1}{4000} \sum_{i=1}^{N} X_i^2 - \prod_{i=1}^{N} \cos(\frac{X_i}{\sqrt{i}}) + 1 \right\}$$

for $|X_i| \le 600.$ (24)

For comparison, SSA, WOA and SSWOA are simulated by utilizing the mentioned four test functions in 3D dimensions. The fitness or cost estimation of the result is characterized as the calculated F(X). The deviation curve of the cost value employing several iterations is revealed in Figs. 3–6.

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Fig. 3. The sphere function's average fitness value by way of diverse algorithms.



Fig. 4. The Griewank function's average fitness value employing diverse algorithms.



Fig. 5. The Rosenbrock function and its average fitness value employing several algorithms.



Fig. 6. Rastrigin function and itsaverage fitness value utilizing several algorithms.

For these algorithms, the population size is alike. Employing different kinds of algorithms, the average fitness value of the sphere function is revealed in Fig. 3. The cost values of the SSA, WOA, and SSWOA algorithms steadily decrease by employing raising iterations. As per the illustration, it can be detected and designed that hybrid SSWOA algorithm is better at 20.408%, 28.57%, 66.39%, 76.65%, 90.1%, 91.65% when compared with standard SSA and 19.58%, 24.99%, 66.39%, 76.65%, 99.01%, 91.65% when compared with standard WOA with the number of iterations.

The Griewank function's average fitness value with many iterations is illustrated in Fig. 4. It is understood from Fig. 4 that, when compared with standard SSA, the suggested hybrid method has lower fitness of 16.16%, 80.00%, 93.32%, 94.22%, 99.52%, 96.83%. While compared with WOA, the adopted hybrid method has lower fitness of 11.70%, 79.93%, 90.9%, 92.34%, 92.48%, and 90.34% with rising iteration. The simulated results show that the suggested hybrid method is superior to the single algorithm and has a much better balance between exploitation and exploration ability.

The Rosenbrock function's average fitness value is displayed in Fig. 5 by making use of many algorithms. The projected technique has a healthier fitness value of 13.04%, 81.28%, 90.44%, 99.6%, 98.87%, 98.53% and 9.09%, 77.69%, 87.78%, 93.16%, 97.06%, 1.10% as compared with standard SSA and WOA. Figure 6 shows the average fitness value of the Rastrigin function. While we compare it by standard SSA and WOA, the suggested method evinces superior fitness value of 41.49%, 45.16%, 58.16%, 73.14%, 94.09%, 91.68% and 37.94%, 40.48%, 55.99%, 69.68%, 93.135%, 89.14%.

Figure 7 illustrates the antenna's gain curve across various frequencies when employing different algorithms. Meanwhile, Fig. 8 showcases the reflection coefficient s_{11} under diverse phase angles, each assessed through various algorithms. Based on these graphical representations, it becomes evident that the proposed hybrid algorithm outperforms both SSA and WOA.

From the obtained results, it is clearly shown that the hybrid algorithm proposed has more energetic investigating capability and quicker convergence speed when it is compared with the standard of the SSA or the WOA. Figures 9 and 10 clearly show how the excitation amplitude and phase of the element are obtained by using SSWOA.



Fig. 7. The performance comparisons of the pattern of gain utilizing diverse algorithms.



Fig. 8. The performance comparisons of the reflection coefficient S_{11} (dB) utilizing diverse algorithms.

Author	Algorithms	Iterations	Fitness	Angle
Sharaqa and Dib [4]	Self-Adaptive Differential	800	NR	300
	Evolution (SADE)			
Sharaqa and Dib [4]	Biogeography-Based	800	NR	300
	Optimization (BBO)			
Sharaqa and Dib [4]	Firefly Algorithm (FA)	800	NR	300
Guney et al. [8]	Backtracking Search (BSA)	1400	NR	300
Khodier [12]	Cuckoo	300	10	300
Proposed	SSWOA	1200	800	NR

Table 1: Comparison table from existing works

NR-Not Reported



Fig. 9. Amplitude excitation using SSWOA.



Fig. 10. Phase excitation using SSWOA.

IV. CONCLUSION

It may be stated that a hybrid SSWOA optimization algorithm can be suggested for the radiation pattern syn-

thesis of EEA. The above-suggested method only needs simple mathematical processing. Moreover, it is effective to put into practice. In EEA pattern synthesis, the numerical results do demonstrate that the suggested algorithm is efficient and achieves the best possible design with high accuracy as well as fast convergence swiftness. Further, the results show that the suggested technique performs better when it comes to the sphere, Rastrigin, Griewank, Rosenbrock, and test functions. The method proposed clearly shows a superior excitation amplitude and phase for a varying number of elements.

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