



## SLL reduction in Antenna array using modified PSOGSA algorithm

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### Abstract

*In this work, we used a new hybrid population-based algorithm (PSOGSA) which is a combination of Particle Swarm Optimization (PSO) and Gravitational Search Algorithm (GSA). It can be used to improve the side lobe level of a linear array of antennas for use in personal communication devices in order to satisfy the situation of huge demand and compressed development cycle for antenna design. Using PSOGSA we can optimize the spacing between the elements of the linear array to produce a radiation pattern with minimum side lobe level and null placement control*

### 1.1 Introduction

The concept of an antenna array was first presented in military applications in the 1940s. This growth was significant in wireless communications as it enhanced the reception and transmission patterns of antennas used in these schemes. The array also enabled the antenna scheme to be electronically steered to receive or transmit information mainly from a specific direction without automatically moving the structure. As the field of signal processing developed, arrays could be used to receive energy (or information) from a specific direction while refusing information or nulling out the energy in unwanted guidelines. Consequently, arrays could be used to mitigate intentional interference (jamming) or unintentional interference (radiation from additional sources not meant for the scheme in question) directed toward the communication system. [1]

An antenna is a transducer to transform from a high frequency electric current to radio waves and vice versa. An antenna is used to transmit and receive radio waves. In transmission, a radio transmitter delivers an oscillating radio frequency electric current to the antenna's terminuses, and the antenna emits the energy from the current as electromagnetic waves (radio waves). [2]

### 1.2 Antenna array

An antenna Array is a configuration of individual radiating components that are organized in space and can be used to yield a directional radiation pattern. Single-element antennas have radiation patterns that are wide and hence have a low directivity that is not appropriate for long distance communications. A high directivity can be still be attained with single-element antennas by increasing the electrical sizes (in terms of wavelength) and hence the physical size of the antenna. Antenna arrays come in numerous geometrical shapes, the most common being; linear arrays (1D). Arrays generally employ identical antenna elements. The radiating pattern of the array depends on the shape, the distance among the elements, the amplitude and phase excitation of the elements, and also the radiation pattern of discrete elements. [3]

The location of the  $n$ th antenna element is labelled by the vector  $d_n$ , where

$$d_n = [x_n y_n z_n]$$



## 1.3 Types of Antenna Arrays

### 1.3.1 Active and Passive array

An active array is one in which an active element (oscillator, amplifier, or mixer) is connected to the path of each radiator. These elements, along with the radiator, form the array module. Active antenna arrays are categorized as receiving, transmitting, and transceiving. Active antenna array advantages include the capability to increase radiated power, decrease thermal losses, increase reliability, and reduce the length of the paths between radiators and transceiving circuits.[4]

A passive array is one in which all elements are excited from a common oscillator or connected to a common receiver. Therefore, an immutable part of a passive array is the feed network connecting the elements. Passive antenna arrays are categorized as receiving, transmitting, and transceiving. They are finding wide use in variable purpose radars.

### 1.3.2 Linear Array

A linear array is one consisting of a group of identical elements placed in one dimension along a given direction. Linear arrays may have equidistant or non-equidistant element spacing's. They are used in the analysis of the directional properties of arrays in antenna theory, and as building blocks for forming an array of arrays.[5]

### 1.3.3 Planar Array

A planar array has all elements located in a single plane occupying a definite area. Planar arrays have different configurations of elements. Rectangular triangular, or hexagonal, in which the elements are positioned at the vertices of the rectangles, right triangles, regular hexagons and also at the center of the hexagon.

### 1.3.4 Adaptive array

An adaptive array consists of an N-element array (usually in the receiving mode), where the useful signal is maximized based on an analysis of the signal to interference ratio. An important

aspect of adaptive arrays is the appropriate choice of weighting coefficients  $W(t)$ , which are placed between the antenna elements and a combining network. In the general case, the vector  $W(t)$  must have the capability of changing the amplitude and phase of the received signal from each element.[6]

### 1.3.4 Cylindrical Array

A cylindrical array is one whose radiators are positioned on a cylindrical surface. The radiators used are wire and slot dipoles, open ended waveguides and horns, and spiral and dielectric rod antennas.

### 1.3.5 Multi-beam Array

A multi-beam array supports the generation of several beams that can be used simultaneously for surveillance of a given sector. Each beam has its corresponding separate input channel. The basic element supporting generation of several beams is the multiple beam forming network.

## 1.4 Modified Particle Swarm Optimization Algorithm

Particle Swarm Optimization (PSO) is a population based stochastic optimization tool inspired by social behavior of bird flock, fish school etc. as developed by Kennedy and Eberhart in 1995. In PSO, a member in the swarm, called a particle, represents a potential solution, which is a point in the search space. The global optimum is regarded as the location of food. Each particle has a fitness value and a velocity to adjust its flying direction according to the best experiences of the swarm in search for the global optimum in the D-dimensional solution space. The steps involved in modified PSO are given below:[6]

### Step 1:

Initialize positions and associate velocity to all particles (potential solutions) in the population randomly in the D-dimension space.

**Step 2:**

Evaluate the fitness value of all particles.

**Step 3:**

Compare the personal best ( $p_{best}$ ) of every particle with its current fitness value. If the current fitness value is better, then assign the current fitness value to  $p_{best}$  and assign the current coordinates to  $p_{best}$  coordinates.

**Step 4:**

Determine the current best fitness value in the whole population and its coordinates. If the current best fitness value is better than global best ( $g_{best}$ ), then assign the current best fitness value to  $g_{best}$  and assign the current coordinates to  $g_{best}$  coordinates.[5]

**Step 5:**

Update velocity ( $V_{id}$ ) and position ( $X_{id}$ ) of the  $d$ -th dimension of the  $i$  –  $th$  particle using the following equations:

$$V_{id}^t = \omega(t) * V_{id}^{t-1} + c_1(t) * rand1_{id}^t * (pbest_{id}^{t-1} - X_{id}^{t-1}) + c_2(t) * (1 - rand1_{id}^t) * (gbest_d^{t-1} - X_{id}^{t-1})$$

$$V_{id}^t > V_{max}^d \text{ or } V_{id}^t < V_{min}^d, \text{ then } V_{id}^t = U(V_{min}^d, V_{max}^d)$$

$$X_{id}^t = rand2_{id}^t * X_{id}^{t-1} + (1 - rand2_{id}^t) * V_{id}^t$$

$c_1(t), c_2(t)$  = time-varying acceleration coefficients with  $c_1(t)$  decreasing linearly from 2.5 to 0.5 and  $c_2(t)$  increasing linearly from 0.5 to 2.5 over the full range of the search, and  $w(t)$  = time-varying inertia weight changing randomly between  $U(0.4 ; 0.9)$  with iterations,  $rand_1, rand_2$  are uniform random numbers between 0 and 1, having different values in different dimension,  $t$  is the current generation number.

The above equation has been introduced to clamp the velocity along each dimension to uniformly distributed random value between  $V_{min}^d$  and  $V_{max}^d$  if they try to cross the desired

domain of interest. These clipping techniques are sometimes necessary to prevent particles from explosion. The maximum velocity is set to the upper limit of the dynamic range of the search ( $V_{max}^d = X_{max}^d$ ) and the minimum velocity ( $V_{min}^d$ ) is set to ( $X_{min}^d$ ).[7]

However, position-clipping technique is avoided in modified PSO algorithm. Moreover, the fitness function evaluations of errant particles (positions outside the domain of interest) are skipped to improve the speed of the algorithm.[5]

**Step 6:**

Repeat Steps 2 to 5 until a stop criterion is satisfied or a pre-specified number of iteration is completed, usually when there is no further update of best fitness value.[8]

**1.6 Use of PSO GSA in SLL Reduction**

Consider a linear antenna array, with  $2N$  isotropic radiators placed symmetrically along the  $x$ -axis. The array factor in the azimuth plane can be written as,

$$AF(\phi) = 2 \cdot \sum_{n=1}^N I_n \cos[k \cdot x_n \cdot \cos(\phi) + \phi_n]$$

Where  $k$  is the wave number, and  $I_n, \phi_n$ , and  $x_n$  are, respectively excitation magnitude, phase and location of the  $n$ -th element. If we further assume a uniform excitation of amplitude and phase (that is  $I_n = 1$  and  $\phi_n = 0$ ) for all elements), the array factor can be further simplified as: [9]

$$AF(\phi) = 2 \cdot \sum_{n=1}^N I_n \cos[k \cdot x_n \cdot \cos(\phi)]$$

Now the statement of the problem, addressed here, simply reduces to: apply the PSO GSA algorithm to find the locations  $x_n$  of the array elements that will result in an array beam with minimum SLL and, if desired, nulls at specific directions.[5]

For side lobe suppression, the *fitness* function used is:

$$F = \sum_i \frac{1}{\Delta\phi_i} \int_{\phi_{li}}^{\phi_{ui}} |AF(\phi)|^2 d\phi$$

To minimize both of them we use the above fitness function as and apply PSO GSA to it.

### 1.7 Results

Before simulations IWO, GA, TSA, MA, PSO and PSO GSA have several parameters which should be initialized. These parameters have been given below. [10]

For PSO we use the following settings:

- Swarm size = 30,
- $c_1 = 2, c_2 = 2,$
- $w$  is decreased linearly from 0.9 to 0.2,
- maximum iteration = 1000, and
- Stopping criteria = maximum iteration.

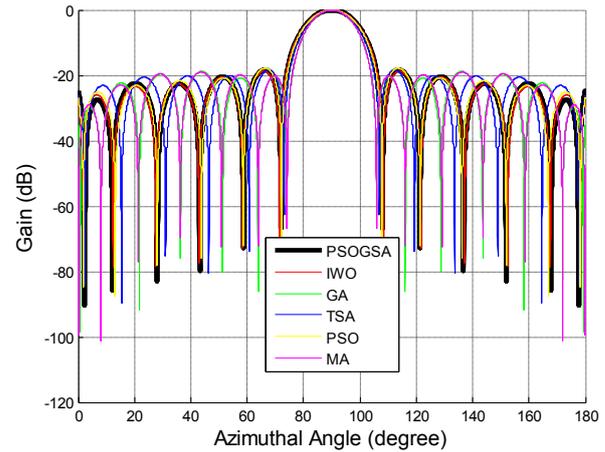
For PSO GSA we use the following settings:

- Population size = 30,
- $c'_1 = 0.5, c'_2 = 1.5,$
- $w$  is any random number in  $[0, 1],$
- Gravitational constant,  $G_0 = 1,$
- $\alpha = 20,$
- maximum iteration = 1000, and
- Stopping criteria = maximum iteration.

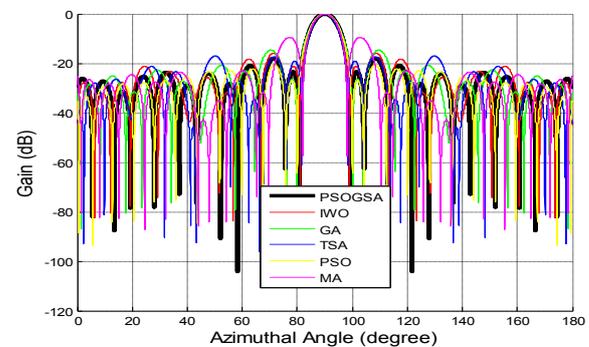
For IWO we use the following settings:

- $sd_{max} = 0.1, sd_{min} = 10^{-5},$
- initial number of plants = 30,
- maximum number of seeds = 5

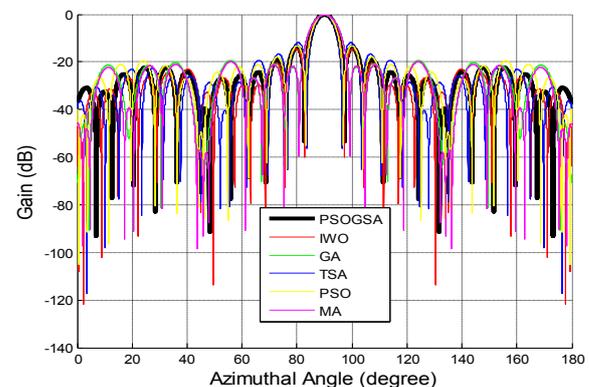
For rest of the competitor algorithms, we used the best possible parametric setup as explained in their respective literatures. All simulation results have been plotted as the Gain versus Azimuth Angle plot.[11]



**Figure 1.1:** Normalized pattern for 12 element array



**Figure 1.2:** Normalized pattern for 12 element array



**Figure 1.3:** Normalized pattern for 12 element array

### Conclusion

All the simulation results show the array pattern from the PSO GSA algorithm using different number of array elements compared with that of the afore mentioned algorithms.

From the results it is clear that PSO-GSA has minimized SLL to the greatest extent and has a low gain value at the null directions as well. Thus, we can say that PSO-GSA has successfully minimized both the sidelobe level and the required null directions. Therefore, we can use PSO-GSA algorithm in the synthesis of linear array geometry for the purpose of suppressing side lobes and null placement in certain directions. PSO-GSA can be used successfully to optimize the locations of array elements to exhibit an array pattern with either suppressed side-lobes, null placement in certain directions

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