Generation of Ramp Pattern using Modified Differential Evolution algorithm

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Abstract: This paper presents a new pattern synthesis for linear arrays using modified differential evolution based on harmony search algorithm as an optimization technique. The main objective is to obtain suitable current excitation amplitude and phase distribution for the linear array antenna elements so that it can produce the desired ramp shaped beam radiation pattern. There are many optimization techniques available for beam shaping but a new algorithm can efficiently control both the main lobe shaping and the number of sidelobes which are present in radiation pattern. The modified differential evolution (MDE) strategy has been developed by adding three control parameter settings into the classical DE. For this purpose a new mutation and crossover strategies are considered. The synthesis of desired ramp shaped beam pattern results have been shown that MDE converges as faster and requires less computation than the other methods.

Index terms: Antenna Arrays, Pattern Synthesis, Control Parameters, Modified Differential Evolution Algorithm, Beam-Former, and Ramp Shaped-Beam Patterns.

I. Introduction:

In modern wireless communication systems, antenna arrays play an important role in detecting and processing signals arriving from different directions. Array pattern synthesis is required in different applications like phased array radar cellular and mobile communication, satellite, radar and military systems for the improvement of signal quality [1].

In recent technology, the need of antenna arrays with shaped beam patterns are increased day by day for the purpose of improvement of communication in which specific beam shape is required for scan and non scan application [2]. In this present work, the objective is to generate a ramp shaped beam radiation pattern from the antenna array pattern synthesis.

This problem consists of finding weights [3] that satisfy a set of specifications and also used to determine the physical layout of the array to produce a desired shaped beam radiation pattern [4]. To obtain this, a set of amplitude excitation coefficients and excitation phases that closely produce desired beam shapes are required. These are carried out by the equally spaced and unequally spaced linear aperiodic arrays. The shape of the desired radiation pattern can vary depending on the application. Ramp shaped beams does not exhibit symmetry about the bore sight direction like pencil beams [5]. These are useful in point to point communication beam shaping pattern can be achieved through amplitude tapering, phase tapering or space tapering or any of the combination of these two. The main aim of this work is to generate a ramp shaped radiation pattern from the antenna arrays using Modified Differential Evolution based on Harmony Search Algorithm.

Many of the antenna array pattern synthesis techniques can be found in the literature for the generation of shaped beam radiation patterns [6]-[7]. G. S. N. Raju, A. Sudhakar et al., [8]-[9] has been showed that generation of optimized ramp type of radiation patterns from an antenna array pattern synthesis. In the above synthesis procedures, both the amplitude and phase of each array element are optimized. G. S. N. Raju, A. Sudhakar et al., [10]-[11] developed and reported the realization for the generation of ramp and stair – step patterns using amplitude control and phase-only control techniques. To avoid the complexity involved in conventional analytical synthesis methods [12].

In recent years, population based stochastic methods such as evolutionary algorithm [13], Genetic algorithm (GA) [14], Simulated annealing (SA) [15] Canonical PSO, some PSO variants has been used for the design of linear array. An adaptive Boolean PSO [15] is developed by Zaharis and Yioultsis for beam-forming on linear antenna array. Liu et al., in [16] suggested a new method for the synthesis of linear array with shaped power pattern. Guney and Basnug [17] used Bacterial Foraging Algorithm (BFA) for interference suppression of linear array. Some other G.K. Mahantietal., antenna array design uses firefly and artificial Bees colony algorithm [18].

Recently the most prominent and efficient differential evolution algorithm [19] has been applied for the design of beam shaping antenna array pattern synthesis [20]. It is vary superior than the other methods.
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Interns of convergence speed and robustness. Different improved variants of original classic DE have been suggested that yields a better result in different situations [21]-[23]. In this paper, a modified differential evolution strategy based on harmony search algorithm [24] is used an optimization technique. This method uses three trail vector generation strategies and three control parameter settings. There are usually three evolutionary operations in classic DE, which are summed up to two operations in the Modified DE. These modifications help to overcome some drawbacks of classic DE. A new mutation and crossover strategies are considered in this proposed algorithm.

The rest of this paper is organized as follows: Section II briefly introduces Modified Differential Evolution Algorithm; Section III describes the mathematical formulation and the related work on MDE. The numerical simulation results are reported in Section IV, and finally the conclusion is given in Section V.

II. Modified Differential Evolution Algorithm:

1. Initialization

Classic DE starts with a new population of NP, D-dimensional vectors $X_{i,G} = [1,2,\ldots,NP]$ where the index ‘i’ denotes the population and ‘G’ denotes the generation to which the population belongs. It involves two stages, namely, initialization and evolution. Initialization generates initial population $P_0$ then the population evolves from one generation ($P_n$) to the next ($P_{n+1}$) generation until the termination conditions are met. While evolving from one generation ($P_n$) to next ($P_{n+1}$) generation DE depends on three main operators, namely, mutation, crossover and selection that are executed in sequence. There are three evolutionary operations in classic DE which are summed up to two operators in the modified differential evolution algorithm [MDE].

2. Differential Mutation and Crossover:

For each individual $X_{i,G}$ in the population, a mutant vector $V_{i,G}$ is produced according to the following formula.

$$V_{i,G} = \begin{cases} X_{i,G} + \sum_{j \neq 1} F \cdot (X_{r_j,G} - X_{r_{j+1},G}) \cdot \text{rand} j [0,1] \leq C_R \\ X_{i,G} \quad \text{otherwise} \end{cases}$$

Where the indexes $r_1, r_2, r_3 \in \{1,2,3 \ldots ,NP\}$ are randomly selected such $r_1 \neq r_2 \neq r_3 \neq i$.

The element of mutant vector $V_{i,G+1}$ is generated by the differential mutation, whenever a randomly generated number between $[0,1]$ is less than or equal to the CR value otherwise, it is equal to the corresponding element of the individual $V_{i,G}$. $F$ is a real and constant factors $\epsilon \in [0,2]$ which controls the amplification of the differential variation $(X_{2,G} - X_{1,G})$. Different values of ‘$\epsilon$’ could lead to differential mutation strategies such as DE/rand/1/bin and DE/best/2/bin.

3. Selection:

The population for the next generation is selected from the individual in current population and its corresponding trail vector according to the following rule.

$$X_{i,G+1} = \begin{cases} V_{i,G+1} \quad \text{if} \quad f(V_{i,G}) < f(X_{i,G}) \\ X_{i,G} \quad \text{otherwise} \end{cases}$$

Where $f(.)$ is the objective function to be minimized. It is to say that if the new vector $X_{i,G}$ produced by differential mutation and crossover operations yields a lower value of the objective function, it would replace the corresponding individual $X_{i,G}$ in the next generation.

In the modified differential evolution strategy, a new parameter $H_k$ is introduced. The element of mutant vector $V_{i,G}$ is generated randomly in the range between $[0,1]$ greater than the specified constant $H_k$. Otherwise, the element is produced by the classic DE. In this case, the probability that each element of mutant vector $V_{i,G}$ is produced in three ways are followed by the schematic structure as shown in the figure 1. The new way to produce mutant elements introduces the random noise in to the population and improves its diversity. At the beginning of the generation, minimum $C_R$ and maximum $F$ are good to search more space, while maximum $C_R$ and minimum $F$ are good to search accurately and escape the local optimum in the final generations.
As similar to the harmony search, the changes of two important parameters $C_R$ & $F$ are improved as follows.

\[
\begin{align*}
\{ C_R (gn) &= C_{R\text{min}} + (C_{R\text{max}} - C_{R\text{min}}) \cdot \frac{gn}{G_{\text{max}}} \\
F (gn) &= F_{\text{max}} \exp (C \cdot gn), C = \ln \left( \frac{F_{\text{min}}}{F_{\text{max}}} \right) / G_{\text{max}} \}
\end{align*}
\]

Where $C_{R\text{min}}, C_{R\text{max}}$ are the minimum and maximum adjusting rate of $C_R$ and $F_{\text{min}}, F_{\text{max}}$ are the minimum and maximum values of $F$ respectively.

To show the excellence performance of Modified Differential Evolution Algorithm the specifications have been selected from the parameter values of harmony search. These rules are governed by some parameters, i.e., a scaling weighted factor ‘$F$’ and a probability $CR$, which control the crossover operator. For this purpose the design specifications for the proposed algorithm have been selected from the control parameter values are $HR=0.9, C_{R\text{max}}=0.95, C_{R\text{min}}=0.2, F_{\text{max}}=0.6, F_{\text{min}}=0.1$ and $G_{\text{max}}=10,000$.

III. Mathematical Formulation

1. Array Methodology:
A linear array is an arrangement of antenna elements placed along a straight line with the orientation of each element in same direction. The elements are spaced at a distance of $\lambda/2$ in order to avoid mutual coupling and grating lobes. The geometry of ‘$N$’ isotropic elements with equal inter-element spacing ‘$d$’ placed in $Z$-direction as shown in figure 2. The array factor in the azimuth plane is given as

\[
E (u) = \sum_{n=1}^{N} A(X_n) \; e^{i(KL \; u \cdot X_n + \phi(X_n))}
\]

$X_n$ is the position of the $n^{th}$ element which is useful for both odd and even number of elements in the array and is given in equation form.
Where $A(x_n)$ and $\phi(x_n)$ are the amplitude and phase distributions of $n^{th}$ element.

$$X_n = \frac{2n-N-1}{N}$$  \hspace{1cm} (5)

‘$k$’ is the wave number $= 2\pi/\lambda$.

‘$\lambda$’ is the operated wave length

‘$L$’ is the length of the array

$u_1 = u-u_0$

$u = \sin \theta$

$u_0 = \sin 0_0$

$0_0$ is the scan angle

2. **Objective Function:**

The goal of the optimization is to generate desired ramp pattern of a specified width ($u_0$) with acceptable sidelobe level by employing non uniform excitations to individual elements of the antenna array. The normalized amplitudes in the search range $[0, 1]$ and progressive phase shift values are in between $-\pi$ to $\pi$ are taken as the optimization parameters. Therefore, the fitness function is given as

$$f = \min (w_1 e_1 + w_2 e_2)$$ \hspace{1cm} (6)

Where $w_1$ and $w_2$ are the controlled weights and sum of the weights should be one that is represented as

$$\sum_{i=1}^{2} W_i = 1$$ \hspace{1cm} (7)

Where ‘$e_1$’ is the mean square error of the main beam region

$$e_1 = \left[ \frac{1}{p} \sum_{i=1}^{p} |E_1(u_i)|^2 \right]^{1/2}$$ \hspace{1cm} (8)

Here ‘$p$’ represents the number of sampling points in the mainlobe region and $E_1(u_i)$ is the error in main beam region and it is calculated as

$$E_1(u) = \{E(u) - F(u); \hspace{0.5cm} 0 \leq u \leq u_0\}$$ \hspace{1cm} (9)

Where $F(u)$ is desired ramp pattern and $E(u)$ is pattern obtained in the evolutionary process.

Therefore the desired ramp pattern is represented by

$$F(u) = \begin{cases} \frac{u}{u_0}; & 0 \leq u \leq u_0 \\ 0; & \text{otherwise} \end{cases}$$ \hspace{1cm} (10)

$e_2$ is the least mean square error in the sidelobe region

$$e_2 = \left[ \frac{1}{Q} \sum_{i=1}^{Q} |E_2(u_i)|^2 \right]^{1/2}$$ \hspace{1cm} (11)

Where ‘$Q$’ is the number of azimuth angles in the sidelobe region and $E_2(u_i)$ is the error obtained in sidelobe region and it is calculated as

$$E_2(u) = \{E(u) - F(u); \hspace{0.5cm} u < 0 \hspace{0.5cm} \& \hspace{0.5cm} u > u_0\}$$ \hspace{1cm} (12)

Where $E(u)$ is the pattern obtained in the evolutionary process and $F(u)$ is desired ramp pattern.

IV. **Numerical Simulation Results**

In order to validate the effectiveness of MDE(HSDEA) for non-convex problems, we first considered a linear array of different elements that are spaced $\lambda/2$ distance apart. For the optimized ramp shaped pattern, the normalized amplitude coefficients and the static phase shifts of the array elements are taken as the optimizing parameters of the MDE algorithm. As can be seen, the average convergence rate for each of the optimized radiation patterns. It should be pointed out that although the same optimality criterion is adopted in the subsection, the specific forms of the fitness functions are different, thus leading to different convergence accuracy.

It is clearly seen that the required number of fitness evaluations varied from 1000 to 60,000 for different cases which takes only about a minute to hours. Figure 3 presented the performance of the behavior of cost function versus the number of iterations. In this paper the resultant optimized ramp patterns are observed for different finite widths ($u_0$) of 0.2, 0.4 and 0.6 by varying the number of elements. Both the amplitude and phase excitation coefficient weights of each element is optimized to generate the desired radiation pattern. Radiation patterns are briefly discussed in the following different cases.
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Case 1: Ramp patterns are generated at finite width \(u_0\) 0.2 for different elements. Figure 4 shows the radiation pattern for 60 elements and its corresponding amplitude, phase distributions are presented in figure 5, figure 6. Radiation pattern obtained for 100 elements is reported in figure 7 its corresponding amplitude distribution and phase distribution for 100 elements are presented in figure 8 and figure 9.

Case 2: Ramp patterns are generated at finite width \(u_0\) 0.4 for different elements. Figure 10 represents the radiation pattern for 60 elements, figure 11 and figure 12 reported its amplitude and phase distributions. Finally for 100 elements radiation pattern are reported in figure 13 and corresponding amplitude, phase plots are given in figure 14 and figure 15.

Case 3: Ramp patterns are obtained at finite width \(u_0\) 0.6 for different elements. Figure 16 shows the pattern for 60 elements, figure 17 and figure 18 reported its amplitude and phase distribution plots. Similarly for 100 elements radiation patterns are reported in figure 19 and its amplitude, phase patterns are presented in figure 20 and figure 21.

As can be seen from these resultant patterns finally modified differential evolution based on harmony search algorithm has a more robust exploration ability to reach the optimal point in the search space. Thus it shows that the method has good ability to achieve the global minima for a non-convex problem. As can be seen from this simulation results, the DES family are successful in reaching the optimal value with in 100,000 number of fitness evaluations (NOFE) while the GA and PSO method failed and DES requires the least NOFE among various methods for same accuracy. Finally the best result optimized ramp patterns are observed for different elements at different widths with this new algorithm.
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Figure 5: Amplitude distribution for 60 elements at 0.2 width

Figure 6: Phase distribution for 60 elements at 0.2 width

Figure 7: Radiation pattern for 100 elements at 0.2 width
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Figure 8: Amplitude distribution for 100 elements at 0.2 width

Figure 9: Phase distribution for 100 elements at 0.2 width

Figure 10: Radiation pattern for 60 elements at 0.4 width
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Figure 11: Amplitude distribution for 60 elements at 0.4 width

Figure 12: Phase distribution for 60 elements at 0.4 width

Figure 13: Radiation pattern for 100 elements at 0.4 width
Figure 14: Amplitude distribution for 100 elements at 0.4 width

Figure 15: Phase distribution for 100 elements at 0.4 width

Figure 16: Radiation pattern for 60 elements at 0.6 width
Figure 17: Amplitude distribution for 60 elements at 0.6 width

Figure 18: Phase distribution for 60 elements at 0.6 width

Figure 19: Radiation pattern for 100 elements at 0.6 width
A new algorithm Modified Differential Evolution based on Harmony Search Algorithm is applied to the pattern synthesis of linear array elements that are spaced \( \lambda/2 \) distance apart a ramp shaped beam radiation patterns are observed. It is obvious from the results presented, that the optimized results are more close to the desired radiation patterns. In this paper the new optimization technique combines the advantages of both harmony search algorithm and differential evolution algorithm the synthesis results showed that the Modified DE converges faster and requires less computation that the other methods.

References

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