



Modified Caiwo Based on Dynamic Range Ratio forAntennaArrays

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Abstract

Due to the high directed radiation patterns required for mobile communication andspatial detection, planar circular antenna arrays require pattern synthesis, which is crucial.For this implementation, IWO was employed. The major motive is to determine if employingCAIWO and dynamic range ratio trade off enhances results in side-lobe reduction while restricting mutual coupling. by thoughtfully selecting a dynamic factor in the range of 20 to30 dB. The investigation of the impact of chaotic characteristics on sinusoidal mapping for chaos in antenna design is the main objective of the study. By the use of MATLAB programming, the separation in between the patterns of antenna's radiation and the number of antenna elements taken into account are analyzed. It has been demonstrated that changing the chaos factor creates times of convergence of chaotic, whichare neededfor attaining optimization's aim of balancing the need for exploitation and exploration, as well as changes then on-uniform planar antennas' radiation factor.

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Keywords:Radiation pattern, tradeoffs,CAIWO, Dynamic IWO, Radiation factor

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1. Introduction

Standarddeviationcanbeemployedi nlogisticchaoticmappingforIWOasitinfluen ces performance. The advantages of the enhanced chaotic weed algorithm are fasterconvergence and more accuracy [1]. IWO has a good level of search of global efficiency. Itscapacity for explorative method is insufficient. The best individual in the room is also notbeing used [2]. The locative of enough global coverage to the extra search area is erroneoussince the rand parameter of the weed algorithm decreases its effectiveness [3]. There is no exact method for enhancing the extraction procedure[4]. Nothing in the algorithm's

implementation changes until the termination condition is satisfied. Lastly, accuracy and speed suffer from an inability to find the right balance between exploration and exploitation. Moreover, there is a lack of demographic diversity and early as simulation [5]. Basically, the chaostheory is used to analyse how specific characteristics behave when they reach a critical situation. Chaostic search can get the majority of data for a specified area.Thegeographicdiversityofthepopulationisincreasedasaresult.Theseearchspacemaynowbeexploredtoagreaterextent.Theanswers wouldthenbedispersedgloballyinanequalmanner.Findingthebestsolutionismadeeasywiththisapproach[6].Theresearcherschooseto



use an initial chaotic population that is created a trandom in order to develop a more diversified seed population [7] as opposed to starting with a seed population of weeds. Also, it is ensured that all of the search territory is covered by the population. A nequation [8] is instead used to identify the best site. These chaotic maps are there as on why the parameter of SD [9], for which a large result is predicted to start the technique, has a small value at its end. A high value at the beginning facilitates exploration. As optimal global is approaching, it is required to choose a parameter value as possible as slow in order to move forward with extracting at this time. Exploration and financial gain are both possible, according to chaos theory [10].

The purpose of this study is to enhance the MATLAB-based CAIWO method to make it more appropriate analysis for antenna [11]. The parts that follow are: Methodology, related work, considerations of design, and then Findings & Conclusion. Concerns with electromagnetic fields have been addressed via the optimization of invasive weeds [12]. It is suffering from erroneous on undesired convergent in multimodal fitness landscapes since it is a stochastic searching mechanism [13]. Based on chaos theory, CAIWO has been implemented to get around these issues. It imitates the invasion and spread of weeds [14] through IWO. The four main phases are initialize, reproduce, geographic disperse, and the exclusion of competitive. There is no balance between the exploration as well as exploitation at IWO since seed production there is depends on a preset distribution of geographic. By using a map of sinusoidal is opposed to a logistic or tent map improves IWO's performance [15].

2. Literature Survey:

D. Li et al., [16] Three distinct forms of particle swarm optimization have been used to boost the performance of the augmented SVM model (PSO). Examples of these include classic PSO, PSO with the

constriction factor, as well as PSO with coefficients of time-varying acceleration (K-PSO). In addition, PRBS-based temporal domain reflectometry may be used to provide a trustworthy dataset for the SVM classifier (TDR).

Tripathi, S., et al. [17] The performance of the single-phase system as a whole requires upgrades due to traction substation control issues.

BS Dhaliwal and others [18] It is necessary to minimize multiple sample benchmark functions using various optimization techniques in order to evaluate how effectively CAIWO works. Numerical findings demonstrate that the proposed method significantly increases algorithm convergence speed and exploration capabilities. There is a statistically significant difference between the design outputs of CAIWO and those of other modern metaheuristics.

Iyampalam, P., et al. [19] This research aims to analyze the Piezoelectric energy harvesting technique. Following the completion of a literature study, several modelling approaches are investigated, contrasted, and the optimal cantilever beam shape as well as chosen materials.

Y. B. Thakare et al., [20] the authors of this work suggest using an IBIWO (Improved Binary IWO) to create thin circular array patterns. A array of circular with the number of elements as well as rings and at least $(\lambda/2)$ -wavelength spacing, equally stimulation, and half-thickness thinning has an HPBW of 50% or below. To demonstrate the effectiveness of the proposed model for concentric circular arrays and simulation outputs are compared to the model's readings [20].

Alghrairiet.al., [21] This strategy consider into account the entire output of the system power as well as efficiency in addition to the number of turns and the size of the coils in the system.

Varamini, G., et al. [22] Later in the classification phase, SVM algorithms



are utilized to detect these properties. The assessment findings show that these techniques

enhance performance in noisy conditions while handling large amounts of data. [22]

M. M. Jawad et al. [23] The electromagnetic (EM) field has made great use of the efficient optimization method known as invasive weed. The proposed method is assessed against existing findings using simulations of IBIWO's suggested thinning arrays. [23]

Pranowo, N. Sharma, and co. Use of the intelligent monitoring system based on Node MCU presented in this article is one technique to monitor health of the machine and notify users when something goes negative. [24].

The creation of a new maintenance plan will be aided by maintaining the real-time data of the history. [25]

Saikumar K. et al. [26] A novel hybrid CNN approach may be used to minimize the side-lobe level of huge concentrated CCA with high side-lobe levels. The issue of optimizing the continuous excitation current may be solved with CSIWA. [26]

3. Methodology

It is important to notify that Holland developed the GA-genetic algorithm as the first instance of natural heuristic variety in 1975. Dorigo proposed the idea of ant colony optimization in 1996. Kennedy & Eberhart first proposed PSO in 1995. Bastruck & Karaboga are credited with developing the Artificial Bee Colony algorithm in 2006.

IWO makes the assumption that weeds are densely produced, which benefits valuable plants. Sigma front features of marijuana include steadiness under challenging circumstances. The ideal mathematical function is found using a meta-heuristic technique developed by Xing and Gao that uses compatibility and unpredictability. A deterministic method can describe the chaotic behavior of dynamic systems with nonlinear nature that are too sensitive to the beginning circumstances. Its implications can help us understand many evolutionary processes.

Table: 1 Deep survey on antenna design

Method	Reference
Power estimation in modern antenna system	Hoang et al., 2022 [16]
A multi grid power balancing system using PSO	Wirasanti et al., 2021 [17]
A single core current balancing and fault detection	Naphade et al., 2021 [18]
A reactive power adjustment using GA	Molina-Viloria et al., 2021 [19]
A critical insight energy differentiation	SM et al., 2021 [20]
An advanced CPW antenna system	Alghairi et al., 2022 [21]
An accurate low energy antenna design for 4G application	Rao et al., 2021 [22]
Typical CPW: Fed antenna a design for low power circuits	Liu et al., 2014 [23]

Ergodicity aids in speed whereas the dynamic quality of chaos aids in the capacity of searching. It is important to

keep in mind that while exploitation is needed to find the most promising solution, exploration provides the



algorithm that identifies the necessary search region. As discovered by metaheuristic evolutionary approaches, chaos is ideal for limited nonlinear systems with ergodic/stochastic features. The map of the sinusoidal is chosen for the required enhancement based on the discussions. For the map, the

sinusoidal sequence is given by $\chi_{k+1} = a(1 - \chi_k)$ for the space $[0, 1]$.

It is important to keep in mind that the CAIWO's success may be impacted by the variance of the chaotic factor. The shift in patterns is seen in Figures 1 through 10.

Before using the chaotic map, the variables are normalised in the range (0, 1). The X variable is changed into

$$X^- = \frac{X - X_{min}}{X_{max} - X_{min}}$$

A new value, X^{-1} , is obtained through transformation using the sinusoidal sequence. Following that, X^{-1} returns to the range (X_{max}, X_{min}) by

$$x^{-1} = x_{min} + x^- (x_{max} - x_{min})$$

Using a sinusoidal map, chaotic distribution is performed. The second phase involves classifying and ranking each weed according to its suitability for the colony. The final phase involves adding up the number of current generation SD of each weed according to its position in the colony as determined by using table 1.

$$\sigma_j = \sigma_{current} * \frac{j}{P_{sum}}$$

Afterwards, adaptive SD is used to disperse new seeds at random. In the fourth stage, new seeds are flocked to the vicinity of the parent new weed using a sinusoidal map. If these seeds are distributed randomly, a higher fitness rating, the better seeds survive. If not, the hapazard seeds will go on.

4. Flowchart of CAIWO:

This method relies on the dominance of local search to hasten convergence. Rating the seeds is the next step. As a result, people who are less fit are less likely to reach the highest seeds p_{sum} . Step 3 is repeated until the maximum number of iterations is reached or the final step's criteria are satisfied, whichever comes first.

$$X_{k+1} = AX_k^2 \sin(\pi X_k)$$

Chaos sets in when $A=2.3$. Adaptive Dispersion and the operation of system. The weeds with the highest fit should attain the lowest SD, while the weeds with the lowest fitness should reach the largest SD.

j = A list of weeds organized by fitness

P_{sum} = The total number of weeds in the current generation

σ_j = SD of j^{th} weed's seed production.

Solutions are seeded and randomly dispersed throughout the n-dimensional search space. SD is provided by



$$\delta_{cur} = \frac{(iter_{max} - iter)^n}{iter_{max}^n} (\delta_{initial} - \delta_{final}) + \delta_{final}$$

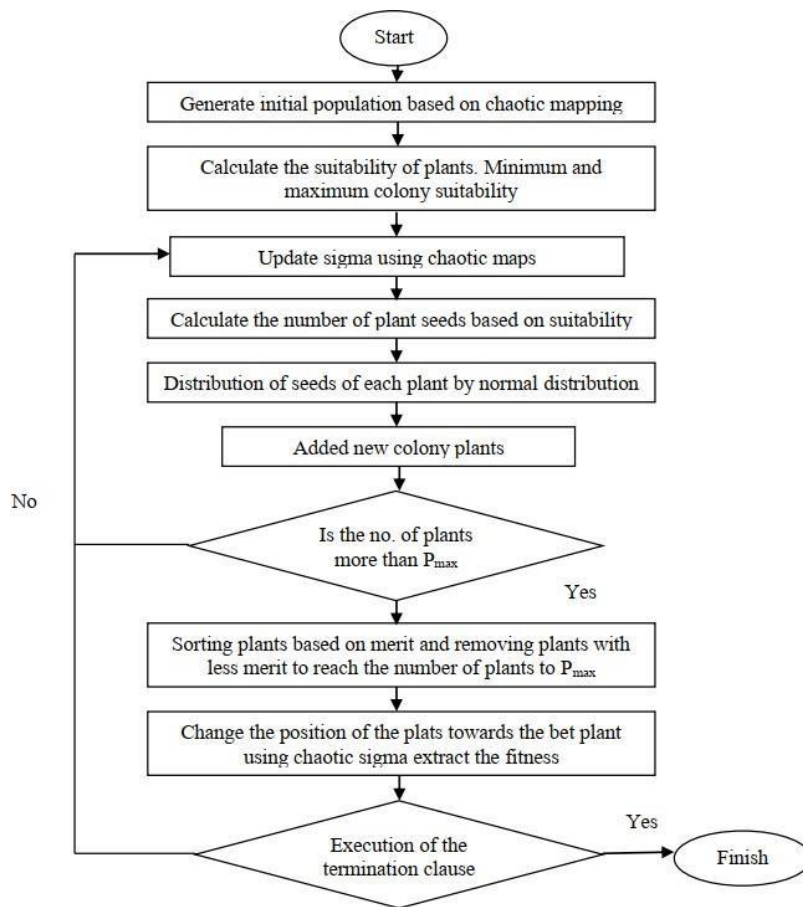


Fig:workFlowoftheCAIWO

With a large SD, the algorithm runs. For the global optimal solution, the search will only be requested to neighbourhoods surrounding the local iteration number increases and the SD value decreases.

New seed:

$$X_{son} = X_{parent} + SD = X_{parent} + radn(0,1) * \delta_{cur}$$

Aresearchprojectonplannerantennaarrays

Step-1:Chaoticmappingbasedoninitialpopulation.

Step-2:MinimumandMaximumarrayestimation

$$\psi_n = -k_r \cos \cos (\phi_0 - \phi_n)$$

Step-3:Chaoticmapsdependingonsigmaelements

$$F_{SLA} + \omega_3 * F_{MSL}$$

Step-4:Distributionofnormaldistributionusingp-max

$$\sum_i^M (0.5 - d_i) d_i < 0.5$$

Step-5:Pointing numberofplants morethanPmax



$$|AF(\phi_{NULL1})| + |AF(\phi_{NULL2})|$$

Step-6: Estimating plants with less merit to reach the number of plants to Pmax

$$F_{MSL} = |AF(\phi_{MSL1})| + |AF(\phi_{MSL2})|$$

Step-7: Chaotic sigma extracts the fitness for power optimization. Step-8: Finish.

5. Discussion on Results:

Since they may create steerable beams that have greater directivity, arrays of antenna are characterized by better qualities as well as increased the efficiency of spectrum. Narrow

bandwidth and low SLL antenna arrays are better able to control interference nulls. To avoid interference, other systems' SLL needs to be maintained to a minimum. Study of array patterns for various levels of chaos

Observation 1:

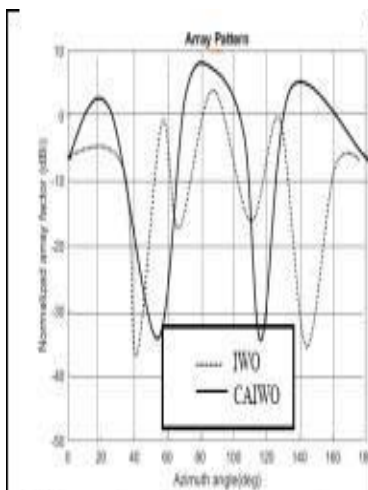


Figure 2: radiation pattern for nine elements with spacing of 0.75λ & chaotic factor value of 2.3 as the critical point

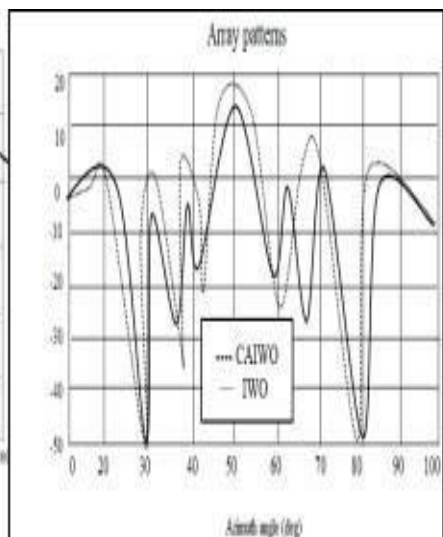


Figure 3: The chaotic factor is 2.0 for nine elements and spacing of 0.75λ .

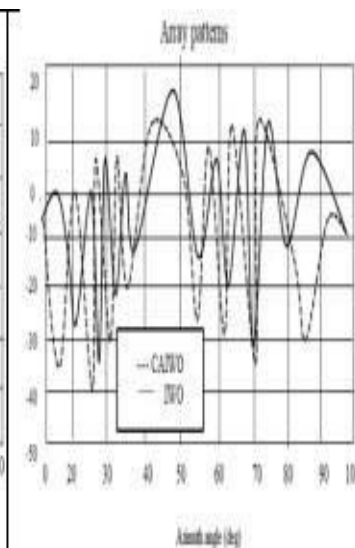
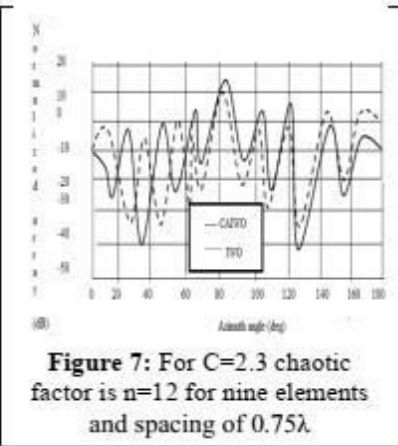
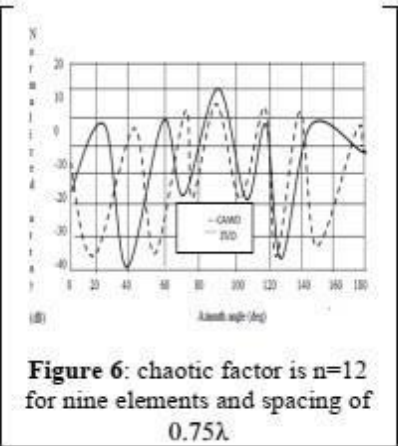
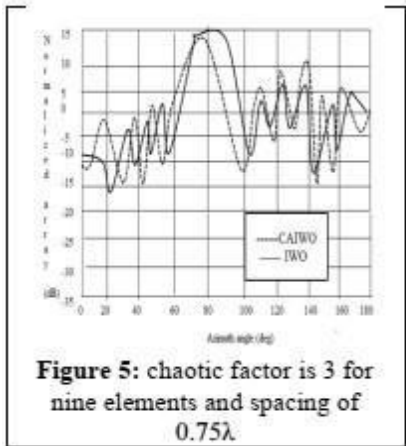


Figure 4: chaotic factor is 3 for nine elements and spacing of 0.75λ

Although 2.3 is the ideal number for c , changing it still produces decent results. Figure 2 illustrates the 4G frequency, $N=9$ array elements, and 0.75λ element spacing. A nine-element spacing antenna is designed in this chaotic environment using an array pattern, as shown in the following figure 3. The nine aspects of chaotic factor spacing are explained in detail in figure 4 above.

When CAIWO and IWO are compared, the suggested design performs better. $C=1$ in Figure 5 refers to the convergence phase. A period of convergence is also seen in this image where $C=1.5$. The periodicity begins in figure 4 with $C=2.0$. In this illustration, $C=5$ and the chaos factor is completely above 3, which corresponds to the chaotic phase.

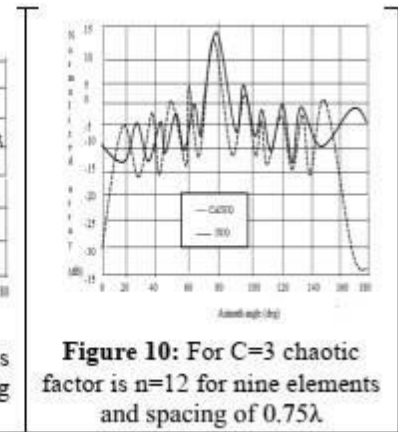
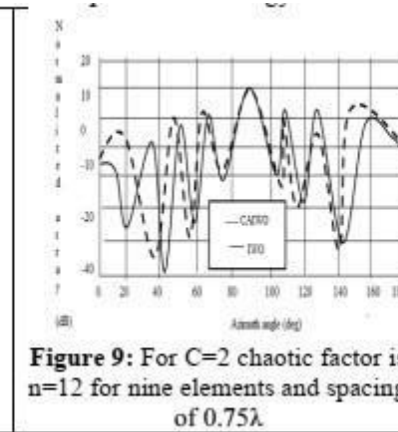
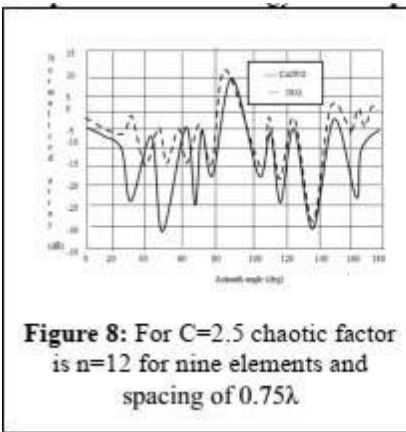




5.2 Observation-

2: Element spacing of 0.75λ for $N=12$. The layout of the 12 element, 9 element

array is explained in figure 6 above. This suggested design outperforms methods and is competitive with current technology.



The $c=2.3$ and $c=2.5$ factors of chaotic are explained in the aforementioned figures 7 and 8. When $n=12$ elements array factors design is compared to CAIWO, IWO technique achieves a fair improvement. The map enters a disorder phase before to the crucial point at chaos factor 2.3, which starts with a chaotic factor 2 phase of periodicity. In figure 8, the phase of the convergence corresponds to the factor of chaos is 1.5. The chaotic element is crucial in Figure 9 and is nearly about to breach the preciosity line. Chaos is 2.5 in Figure 8, which is barely over the critical point. The convergence-related chaos factor 2 is shown in figure 8. Figure 10 clearly explains the important places in relation to the crossing of $C=3$.

6. Dynamic IWO:

The best array pattern has broad or deep nulls in certain interference directions with low side lobe levels. To have an amplitude excitation with a low dynamic range ratio, which aids in the design of the beam forming feed, is the goal of the synthesis. This type of synthesis is carried out under restrictions such as minimum SLL, fixed initial null beam width, and dynamic range ratios smaller than for a particular threshold in current distribution.

Because to the numerous drawbacks of analytical synthesis approaches, such as their tendency to become trapped in local minima, evolutionary methods based on natural events are being employed more and more to tackle antenna synthesis issues. It should be high



lighted that a low DRR of excitation amplitude is required to accomplish array feeding.

IWO is a potent strategy that may be used instead of conventional Meta heuristic techniques. The remarkable sensitivity and adaptability of the invasive weeds allow them to travel freely over agriculture fields and mature to the blooming stage where they produce new seeds. Starting with random parameters for a linear uniform array, the optimization process begins. With these settings, the primary beam is placed between symmetric broad nulls. There are several instances for finding spatial dispersal models that properly distribute seeds around the parent plant in circumstances of alternation.

The model of spatial dispersion in dynamic IWO uses a mutation process to iteratively calculate standard deviation. It is pertinent to note that number of produced seeds

grows randomly and become new plants. The procedure is divided into:

- a) Initialization
- b) Evaluation
- c) Reproduction

d) Spatial dispersion

e) Mutation

f) Limitation

g) Stopping Criterion

The GA's mutation process is employed in dynamic IWO to determine the spatial dispersion parameters. The starting value of the standard deviation is used in place of the actual value if the initial probability is lower. The value of the mutation probability is selected empirically whereas the actual probability diminishes as per exponential pattern. Due to this decrease, when the cost functions are close to the fitted value, the standard deviation is decreased. The method's dynamic nature allows for linear changes in the ratio of mutation operators during the search process. The colony with the fastest-growing and highest-reading plant population will receive the extra seeds. In this case, population will shrink by competitive exclusion. This process keeps the best population in the colony by removing the rest.

The below table shows that the array's $2N=20$ excitation weights now in effect, with deep null set to 14^0 .



Table 2: Current excitation weights with a deep nullen forced at 140 for the array of 2N=20 items.

Pattern	Current excitation at its best (Normalized)					DRR
	1.0000	0.9701	0.9124	0.8310	0.7314	
Chebyshev						3.50
	0.6203	0.5046	0.3910	0.2855	0.3256	
	1.0000	0.9892	0.9548	0.8870	0.7796	
BA						4.10
	0.6405	0.4897	0.3505	0.2445	0.3106	
	1.0000	0.9945	0.9639	0.8926	0.7808	
BFA						4.00
	0.6399	0.4887	0.3518	0.2480	0.3142	
	0.9693	0.9610	0.9257	0.8528	0.7427	
DIWO						4.20
	0.6093	0.4534	0.3312	0.2335	0.2752	
	0.9681	0.9592	0.9102	0.8421	0.6954	
CAIWO	0.6012	0.4502	0.3291	0.2221	0.2621	4.31

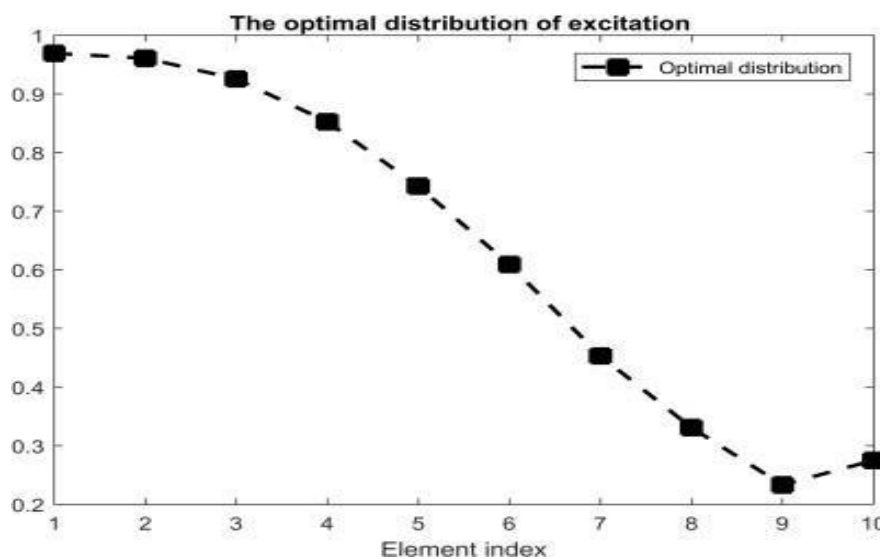


Fig.:The ideal linear array excitation distribution for the situation in Fig.7 (2N=20 with nullen forced at 14°).

7. Conclusion:

For designed antenna array radiation patterns, CAIWO is calculated by the figures of Chaos choice. The radiation pattern analysis has shown that there are many circumstances of convergence, periodicity, as well as chaos depending on the parameter,

ranging from 1 to 3. The sinusoidal map is more successful with a crucial factor of 2.3. In order to effectively balance exploration (Convergence) and exploration (Chaos). The use of MATLAB programming to examine the effects of rising chaos factors on the quantity of antenna components and the distances between their emissions patterns. The major approach



used in this work to investigate invasive weed optimization in compression is modified chaotic adaptive invasive weed optimization. The synthesis of the uniform linear antenna array was successfully completed using a dynamic CAIWO that was provided. The current excitation for each element in the array was determined using this synthesis problem while the phase remained unaltered. There was also consideration for the DRR. The findings shown that, under the constraints of a set main beam width and low DRR, the dynamic CAIWO technique yields normalized array factors with enhanced nulls (deep or broad) in the appropriate directions as well as with the lowest side-lobe levels.

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