

Research Article

Conformal Array Pattern Synthesis Using a Hybrid WARP/2LB-MOPSO Algorithm

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This paper addresses conformal array synthesis as a constrained multiobjective optimization problem. Simultaneous reduction of side lobe level (SLL) and cross-polarization (XPL) level is aimed with a constraint on main beam direction. A hybrid of weighted alternating reverse projection (WARP) and two local best multiobjective particle swarm optimization (2LB-MOPSO) is proposed to optimize the pattern. First, the WARP method finds a moderate and feasible solution. Second, 2LB-MOPSO begins with an initial population including the solution of WARP and penalty functions for constraint handling. Involving WARP result in the initial population of 2LB-MOPSO leads to higher convergence rate, avoiding local extremum traps and less sensitivity to penalty functions. Compared to WARP method which stagnates rapidly, the proposed hybrid method gives better SLL and XPL after adequate iterations. In addition, as 2LB-MOPSO offers a set of optimum solutions (Pareto front) instead of a single solution, this method provides more degrees of freedom in selection of proper practical arrays. Finally, to examine the mutual coupling consideration in array design, the same procedure was applied ignoring the mutual coupling between elements. The results show that the SLL and XPL strongly depend on mutual coupling.

1. Introduction

There is an increasing demand to conformal arrays because of their flexibility in attaching to prescribed shapes of vehicles and aircrafts or ships [1]. Among several conformal arrays, spherical arrays have received considerable attention because of their ability to provide single or multiple beams for full spherical coverage. These antennas have been proposed for applications such as ground station of satellite terminals and telemetry [1, 2]. The main goal in these applications is to obtain pattern with low side lobe level (SLL) and cross-polarization level (XPL).

Unlike planar arrays, radiating elements in the conformal array are oriented in different directions and there is no closed form solution for analysis and consequently synthesis of these arrays [2].

If the array is conformed to curved surfaces, It is not possible to define an array factor. Therefore, unique challenges are posed in synthesis of these arrays. There are several

methods to synthesize a desired pattern in conformal arrays [3–11]. Using high-speed computers, iterative methods are commonly used for pattern synthesis. The least mean squares (LMS) [5–7] and alternating projection [8] are extremely powerful for synthesizing the array pattern. The weighted LMS [9] was introduced where different weights for different directions are required. The weighted alternating reverse projection method (WARP) is similar to weighted LMS method. This method is a combination of weighting LMS and alternating projection. In our recent study, the WARP method has been used for synthesizing of hemispherical conformal array antenna [11]. The results show that WARP starts very well and generates appropriate results after a few iterations. However, it stagnates rapidly. On the contrary, there are evolutionary algorithms which search the solution space more efficiently but they may get trapped in local extremums. To get good results in a reasonable time and to avoid local extremum traps, we can mix the WARP method with an evolutionary algorithm. Some evolutionary

algorithms such as genetic algorithms (GA) [12], simulated annealing [13], and particle swarm optimization (PSO) [14, 15] have been used in pattern synthesis. However, they are single objective techniques and do not consider different objectives (SLL and XPL) separately and hence no trade-off is possible. Therefore, multiobjective (Pareto) optimization techniques which find the best trade-off among conflicting objectives are preferred.

In this paper, a hemispherical conformal array antenna is simulated in CST. This array is composed of rectangular patches conformed to the hemispherical surface. The embedded pattern of each element are calculated by CST and then exported to a MATLAB program where an optimization routine based on hybrid WARP and two local best multiobjective particle swarm optimization (2LB-MOPSO) is used to get the desired pattern with low SLL and XPL. To maintain main beam in a specific direction, a constraint is also added to the optimization which is handled by a constant penalty function. In addition to higher convergence rate and avoiding local extremum traps, involving WARP result in the initial population of 2LB-MOPSO leads to less sensitivity to penalty functions.

Compared to WARP method, the proposed hybrid method improves SLL and XPL about 3 dB and 3.5 dB, respectively. To the best of our knowledge, it is the first time that the SLL and XPL are optimized where the embedded radiation pattern and the mutual coupling effect between elements are taken into account in the optimization process. Furthermore, this hybrid approach provides a set of optimum solutions instead of a single solution in a very short time. This set of solutions gives the designer more degrees of freedom for choosing the final array excitation.

2. Pattern Formula of Conformal Array

The farfield of an arbitrary array can be written in matrix form as follows [1, 9–11]:

$$\mathbf{E} = \mathbf{Y}\mathbf{A}, \quad (1)$$

where \mathbf{E} is the directivity vector $[E(\theta_1, \varphi_1) \cdots E(\theta_k, \varphi_k)]^T$ in k direction $((\theta_i, \varphi_i), i = 1, 2, \dots, k)$, \mathbf{A} is the excitation vector $[a_1 \cdots a_N]^T$ of the array and T denotes the transpose of a matrix. \mathbf{Y} is the geometry-dependent matrix in spherical conformal arrays:

$$Y_{ij} = g_j(\theta_i, \varphi_i) \cdot e^{jk_0[\sin(\theta_i)\cos(\varphi_i)x_j + \sin(\theta_i)\sin(\varphi_i)y_j + \cos(\theta_i)z_j]}, \quad (2)$$

where $x_j, y_j,$ and z_j are positions of the j th element and $g_j(\theta_i, \varphi_i)$ is field-gain of the j th element in the direction of (θ_i, φ_i) , that are not necessarily identical. \mathbf{E} can be a θ or a φ component and \mathbf{Y} is different for different polarizations. Matrices \mathbf{E} and \mathbf{Y} can be written in both polarization components:

$$\mathbf{E} = \begin{bmatrix} \mathbf{E}_\theta \\ \mathbf{E}_\varphi \end{bmatrix}, \quad \mathbf{Y} = \begin{bmatrix} \mathbf{Y}_\theta \\ \mathbf{Y}_\varphi \end{bmatrix}. \quad (3)$$

Therefore, both polarization components of farfield found from (1).

It can be shown [1] that if the mutual coupling of array is considered, then (1) must be modified to

$$\mathbf{Y}_s = \mathbf{Y}(\mathbf{I} + \mathbf{S}), \quad (4)$$

where \mathbf{I} is the identity matrix and \mathbf{S} denotes the S-parameter matrix.

In pattern synthesis of conformal array, we are looking for appropriate excitation vector to get the desired radiation pattern. The excitation values can be found by the least-mean squares (LMS) method minimizing the difference between the realized pattern and the desired pattern, respectively, [1, 3, 4, 9]. In the weighted LMS method, excitation values can be calculated by introducing different weights for different directions [9]. Alternating projection method [16] introduces upper and lower boundaries in pattern and tries to hold the realized pattern within these boundaries. The weighted alternating reverse projection (WARP) method [10, 11] is a combination of weighting and modified alternating projection methods. As was shown in [12], the WARP method is applicable to synthesis of arbitrary array shapes and has better performance, compared to other methods.

3. MOPSO Algorithm

Pareto front in multiobjective contexts plays the same role as “extremum value” in single objective contexts. Since a single optimum solution is meaningless for multiobjective problems, Pareto front corresponds to a set of solutions which present the best trade-off among different objectives.

It is obvious that, the main aim of every multiobjective optimization algorithm is to find the Pareto optimal set. Traditional versions of Pareto optimization algorithms used to define a new aggregative weighted function from individual single objective functions [17]. Since these algorithms check the dominance criteria after each run of the new single objective problem, they tend to be very time-consuming. In addition, these classic approaches tend to be ineffective and inefficient due to their sensitivity to weighting factors [18]. More recent algorithms such as non-dominated sorting genetic algorithms II (NSGAI), multiobjective evolutionary algorithm (MOEA), and MOPSO omitted the “transform to single objective” stage and only use dominance criterion at each iteration. Among several dominance-based multiobjective algorithms, we applied the two lbest MOPSO (2LB-MOPSO) [19] which has been recently introduced as a powerful multiobjective evolutionary algorithm and tested for several standard optimization problems. However, we also compared it with an earlier version of MOPSO proposed by Coello et al. [20] and have been applied for several electromagnetic optimization problems [21]. The results show superiority of 2LB-MOPSO in our array antenna synthesis.

4. Results and Discussions

The proposed method is applicable to arrays on arbitrary surfaces, but in this section the method is applied to synthesis of hemispherical array. Figure 1 shows the conformal array

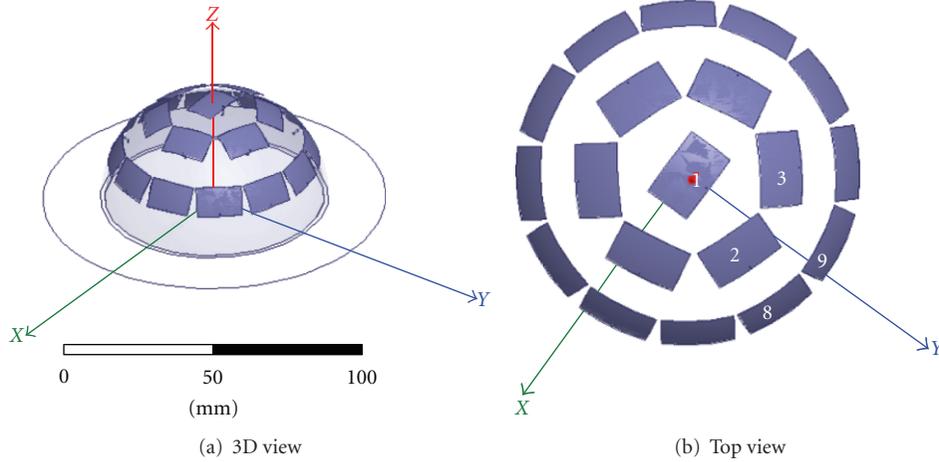


FIGURE 1: Arrangement of a hemispherical conformal array antenna.

structure where the elements are numbered in counterclockwise sense in each row. This array consists of 19 rectangular patches wrapped around a hemisphere with a radius of $1.1\lambda_0$, where λ_0 is the free space wavelength. The operating frequency is 8.3 GHz.

The dimensions of the patches are $10.5 \text{ mm} \times 16 \text{ mm}$. In this model, the patches are mounted on a substrate with thickness of $h = 1.58 \text{ mm}$ and the permittivity of $\epsilon_{r1} = 2.52$. The θ -polarized patches are fed through a coaxial probe. The Coaxial probe is centered in φ -direction, but its position is 4.5 mm far from the patch center in θ direction. The elements around the hemisphere are spaced apart by $0.58\lambda_0$ in the first row from top of hemisphere and $0.5\lambda_0$ in the second row. The rows are spaced apart by $0.58\lambda_0$ from each other with different numbers of elements.

For simplicity, the azimuthal angle has been fixed and the elevation angle has been considered. The embedded-element patterns are evaluated by CST. For the embedded-element pattern extraction, only a single element is fed while the other elements are terminated in match loads. Therefore, they are being excited only by the mutually coupled fields. So, the embedded-pattern carries more information such as the effect of mutual coupling, surrounding elements, and a particular element grid [1, 11]. The embedded-element pattern gives an indication of the performance of the fully excited array and serves as a valuable tool in the array design process.

The third definition of Ludwig is used to define the copolarization E_{copol} and the cross-polarization E_{xpol} [22, 23]:

$$E_{\text{copol}} = E_{\theta} \cos \varphi - E_{\varphi} \sin \varphi, \quad (5)$$

$$E_{\text{xpol}} = E_{\theta} \sin \varphi + E_{\varphi} \cos \varphi. \quad (6)$$

It follows from (5), $|E_{\phi}|$ and $|E_{\theta}|$ are the co- and the cross-polarization components in the $\varphi = 90^\circ$ and $\varphi = 270^\circ$ planes, respectively.

WARP and 2LB-MOPSO are applied to proposed hybrid method for optimization. Once the embedded pattern of elements are obtained, the remaining optimization procedure

will be continued in MATLAB. First, the WARP method is applied to find a preliminary solution. In the second stage of our hybrid algorithm, 2LB-MOPSO searches the vicinity of WARP solution. The considered multiobjective problem is formulated as follows:

$$\begin{aligned} \text{XPL} = f_1 &= \max \left\{ 20 \log \left(\frac{E_{\text{co}}(\theta_{\text{max}})}{E_{\text{xpol}}(\theta)} \right) \right\}, \\ \text{SLL} = f_2 &= \max \left\{ 20 \log \left(\frac{E_{\text{co}}(\theta_{\text{max}})}{E_{\text{co}}(\theta_{\text{SL}_i})} \right) \right\}, \end{aligned} \quad (7)$$

Constraint: $\theta_{\text{max}} = 32$.

Since the pattern synthesis is not a convex problem, the rejection strategy cannot handle the constraint. Therefore, various penalty functions were tested. Generally, evolutionary algorithms are very sensitive to penalty functions and finding a proper penalty is a challenge [24]. However, in proposed hybrid method, adding a good feasible solution (WARP result) reduces this sensitivity whereas a constant penalty function in both SLL and XPL works properly.

In the hybrid method, Initial population of 2LB-MOPSO is random, except a particle that is the result of WARP. First, we compare the 2LB-MOPSO and MOPSO. For both cases the population size is $\text{NPOP} = 50$ and the maximum number of generation is $T_{\text{max}} = 1000$. Number of bins in both directions of objective space in for both 2LB-MOPSO and MOPSO is 20 and the in 2LB-MOPSO maximum count number which is checked to update lbests for each particle is 5. More details of algorithms can be found in [19, 20]. As it is shown in Figure 2, the 2LB-MOPSO overcomes MOPSO. Therefore, the rest of optimizations in this paper are performed using 2LB-MOPSO.

In the following subsections, the proposed algorithm is applied for two case studies of hemispherical array synthesis considering and ignoring the mutual coupling effect.

4.1. Hemispherical Array Synthesis Considering Mutual Coupling Effect. In this section, the embedded pattern which has the information of the effect of mutual coupling has

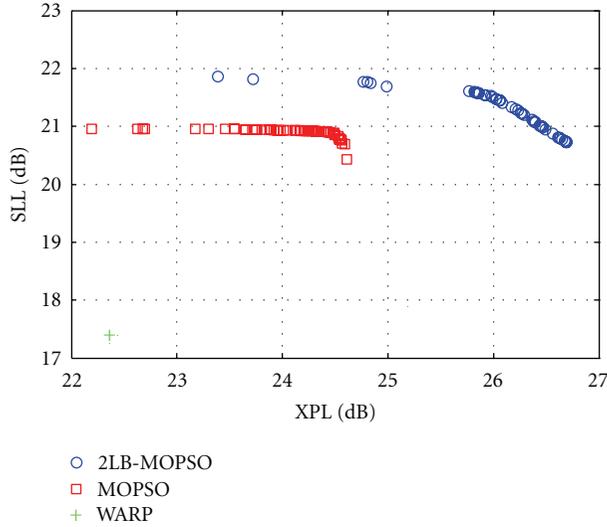


FIGURE 2: Pareto fronts obtained by the 2LB-MOPSO and MOPSO.

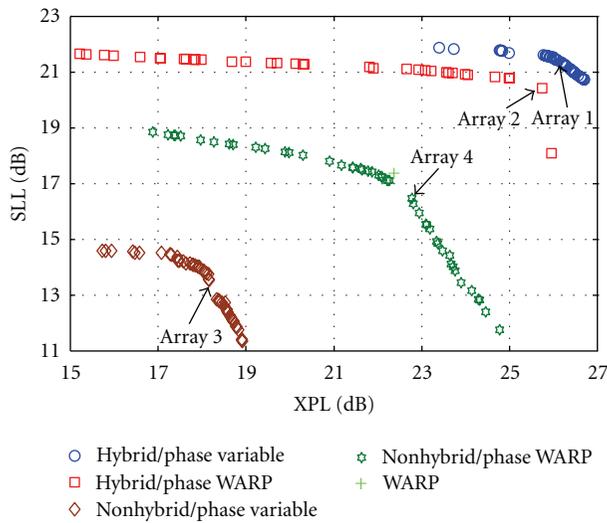


FIGURE 3: Resulted Pareto front of hybrid and nonhybrid, variable phase and WARP phase compared to WARP method result.

been used. To investigate the effectiveness of proposed hybrid method, the hybrid and nonhybrid 2LB-MOPSO were compared. In both of hybrid and nonhybrid scenarios, excitation phase of each elements can be fixed (equal to WARP result phase) or tunable. When phase of excitation is equal to the WARP result phase, the number of variables would be half.

Figure 3 shows Pareto front for 4 cases as follows.

Hybrid/Variable Phase. Uses the amplitude and phases of the WARP method in the initial population and both phase and amplitude of excitations are tunable.

TABLE 1: Comparison of SLL and XPL.

Sample	SLL (dB)	XPL (dB)
WARP	17.5	22.5
Array 1	21.5	26
Array 2	20.5	25.7
Array 3	13.5	18.2
Array 4	16.45	22.8

Nonhybrid/Variable Phase. Does not use WARP solution as an experience and both phase and amplitude of excitations take participates in the optimization.

Hybrid/WARP Phase. Excitation phase of each element is equal to the excitation phase of finding in WARP for total iteration and only amplitudes are optimized. It uses the amplitude and phase of the WARP method in the initial population.

Nonhybrid/WARP Phase. Phase and amplitude of elements are changed during the optimization. It does not use amplitude of WARP result in the initial solution.

As shown in Figure 3, hybrid methods dominate nonhybrid methods. In addition, tuning the phase provides more degrees of freedom and consequently better results. So the best Pareto belongs to hybrid/variable phase method and the worst Pareto belongs to nonhybrid/WARP phase. As a result, If WARP solution uses as an experience, this helps 2LB-MOPSO to obtain better element excitation which results in lower SLL and XPL. Furthermore, only nonhybrid/WARP phase is dominated by WARP result. If nonhybrid method is applied, WARP phase can find better results in comparison with variable phase. To have a numerical comparison, 4 different samples from Pareto fronts are selected (Array 1–Array 4). Corresponding SLL and XPLs are listed in Table 1. From Table 1 obviously Array 1 has about 3 dB and 3.5 dB SLL and XPL, respectively, better than WARP.

Figure 4 compares the co- and cross-polarization patterns obtained by WARP with one of the hybrid and nonhybrid solutions (indicated by Array 1 and Array 3 in Figure 3). As shown in Figure 4, beam width changes in nonhybrid/variable phase and Array 3 solution has higher SLL and XPL than Array 1 and WARP.

The comparison between the proposed hybrid method radiation patterns and the full wave CST simulation for Array 1 is shown in Figure 5 which proves a good agreement between these two results.

4.2. Hemispherical Array Synthesis Neglecting Mutual Coupling Effect. In this section, the mutual couplings between elements are neglected. The farfield of each element are exported from CST with assuming that they are isolated. Then WARP and the proposed hybrid method are applied to this array. Figure 6 shows the resulted Pareto front. It can be seen that the proposed hybrid method at most improves SLL and XPL around 1.5 dB and 0.5 dB, respectively, in comparison with WARP in this situation.

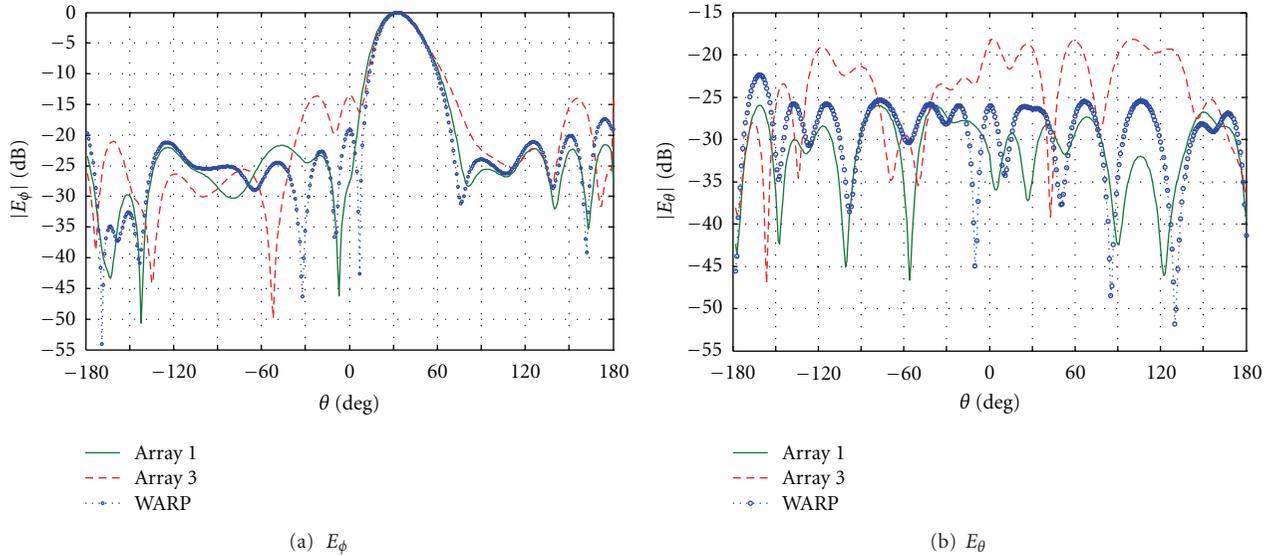


FIGURE 4: Comparison of the realized pattern using hybrid, nonhybrid, and WARP methods in the planes $\varphi = 90$ and $\varphi = 270$.

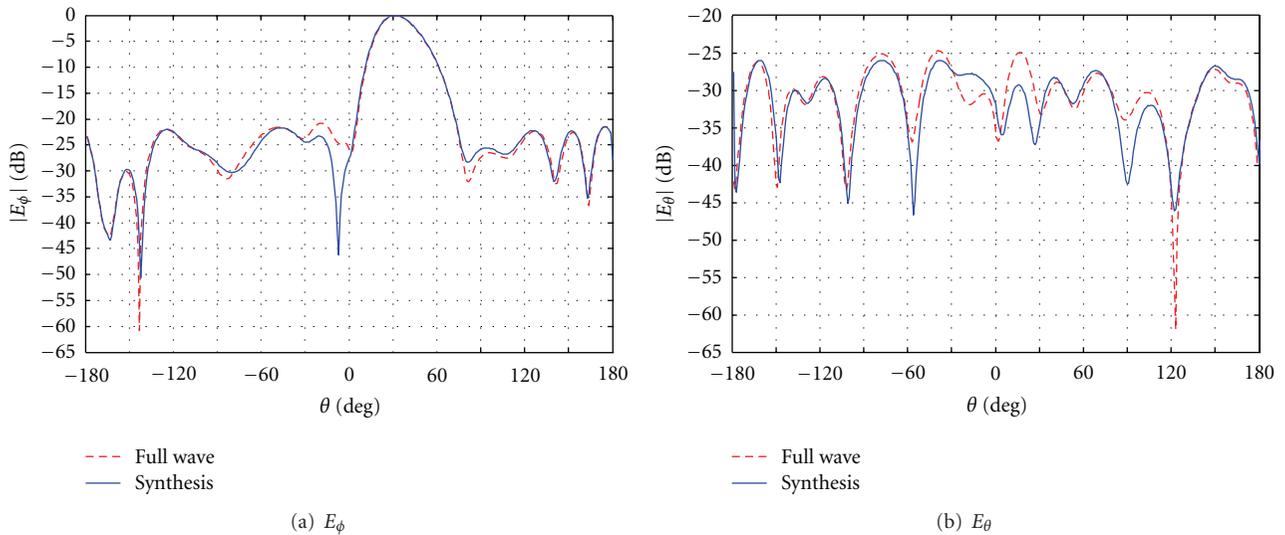


FIGURE 5: A comparison between the realized pattern using the proposed hybrid method and the simulated by CST in the planes $\varphi = 90$ and $\varphi = 270$.

The co- and cross-polarization patterns, one of the hybrid solutions (indicated by Array 5 in Figure 6) that were obtained by MATLAB and full wave, were compared in Figure 7. As it illustrated that the SLL and XPL of full wave simulation changed in comparison with MATLAB result, because the mutual coupling effect in synthesizing of array was not considered in synthesizing by MATLAB.

Also, the mutual coupling in conformal array may be less than planner array, but SLL and XPL are strongly dependent on mutual coupling. Therefore, the mutual coupling plays an important role in synthesis. It is essential to include these effects in the synthesis of array such as conformal arrays.

5. Conclusion

A novel hybrid of 2LB-MOPSO and WARP was proposed and used to optimize the pattern of a hemispherical-conformal array antenna. In the case of known geometry distribution of array elements, the excitation amplitude, and phase of each element were optimized to reduce SLL and XPL, simultaneously.

The results show that the hybrid method leads to better pattern compared to both WARP and 2LB-MOPSO (nonhybrid). In addition, this multiobjective algorithm provides a set of optimum solutions instead of a single solution which helps the designer in his/her final selection. The result of

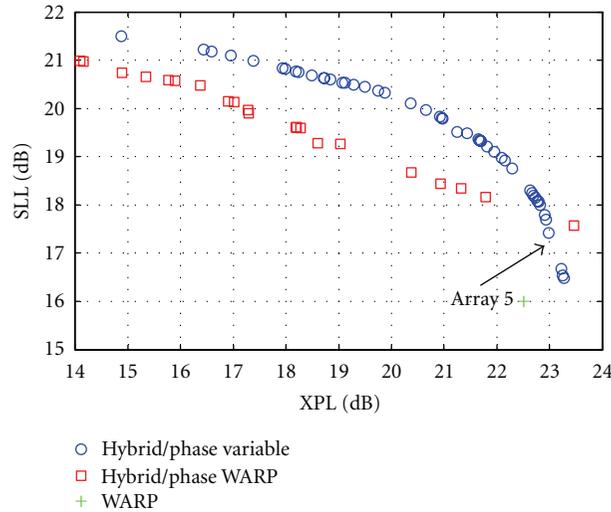


FIGURE 6: Resulted Pareto front of hybrid variable phase compared to WARP with assuming isolated elements.

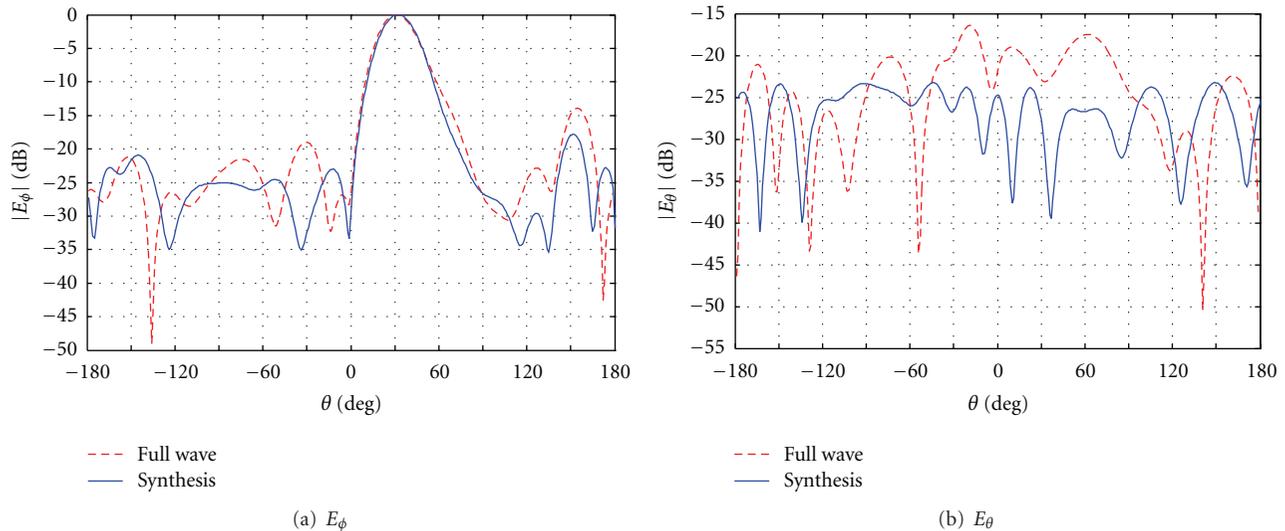


FIGURE 7: A comparison of the realized pattern using the proposed hybrid method and the simulated by CST without considering of the mutual coupling in the planes $\varphi = 90$ and $\varphi = 270$.

the proposed method was verified by full wave simulations. Comparing the results of MATLAB codes and full wave simulations, shows the effect of mutual coupling in the synthesis of hemispherical conformal array should not be neglected because the SLL and XPL are strongly dependent on mutual coupling.

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