

# Neural Network Modeling for the Reduction of Scattering Grating Lobes of Arrays

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**Abstract** – The monostatic radar cross-section (RCS) of an array is seriously deteriorated by the scattering grating lobe. In this paper, the scattering grating lobe of an array is suppressed by metal walls around elements. The artificial neural network with Fourier series-based transfer functions is used to accelerate the design process. A  $1 \times 8$  array with the patch element operating in the range from 9.4 to 10.6 GHz is studied. The monostatic RCS of the array with designed metal walls is compared with that of the array with no metal wall. Simulated results show that the scattering grating lobe of the array with metal walls is suppressed by 5.8 dB at 12 GHz, and the change of radiation performance is acceptable. The design procedure is also available for other arrays with reduced scattering grating lobes.

**Index Terms** – Artificial neural network, metal wall, radar cross-section, scattering grating lobe.

## I. INTRODUCTION

With the development of the wave-absorbing material, the radar cross-section (RCS) of a stealth platform is suppressed effectively. The reduction of RCS of antennas still remains to be challenging work. To reduce the RCS of antennas, some effective methods have been proposed. The diffusing of scattering wave based on the metasurface is a remarkable method [1, 2]. The incident wave is diffused by the metasurface covering on the antenna based on the phase cancellation mechanism. Despite that, the diffusing method is not suitable for wide-angle RCS reduction. The absorbing material is also used to reduce RCS [3, 4]. The radiation performance of the antenna, however, is affected because the material also absorbs the radiated waves. One traditional and effective tool to suppress RCS is the frequency selective surface (FSS) covering on antennas. FSS is made as low-RCS shape to diffuse the outbound waves, and

the incident inbound wave is absorbed by the loaded antenna, whereas the wave in the neighboring band is also able to penetrate FSS because of the imperfect reflection of the outbound wave. The wave with higher frequency than the inbound wave generates noticeable scattering lobes, which highly deteriorate the wide-angle monostatic RCS. It remains a problem for RCS reduction of uniform arrays.

Moreover, the traditional design process is often accompanied by time-consuming full-wave simulations. The artificial neural network (ANN) with the transfer function (TF), as a substitute for the full-wave simulation, has been used for wideband electromagnetic (EM) modeling. By representing the wideband EM response as TF, ANN maps the relationship between the geometric parameters and the response. The trained ANN can obtain the EM response accurately and quickly, and it improves the whole design efficiency.

In this paper, an array whose elements are surrounded by metal walls is proposed for the design of a low RCS array. The array is modeled based on two branch ANNs that independently study the radiation and scattering performances to accelerate the design procedure. In addition, the Fourier series-based TF are used in ANN modeling [5], which has the same order for all samples and fewer coefficient orders than those of the pole-residue-based TF [6], bringing the fast convergency and robustness of the model. The monostatic RCS of the array with no metal wall is also studied for comparison. Simulated results show that the scattering grating lobes are suppressed effectively.

## II. DESIGN OF ARRAY WITH LOW-GRATING LOBES

The scattering wave of a periodic structure with the illuminating of a plane wave contains Floquet-Bloch

modes, as shown in Fig. 1. The Floquet-Bloch waves have transverse wavenumbers as

$$K_s = K_0 \times \sin(\theta) + n \times \frac{2\pi}{P}, n = 0, \pm 1, \pm 2, \dots, \quad (1)$$

where  $K_s$  is the transverse wavenumber of the scattering wave,  $K_0$  is the wavenumber in vacuum,  $\theta$  is the incident angle, and  $P$  is the period of the structure. The transverse wavenumber of the zero-order mode is the same as that of the incident wave. The first high-order mode, i.e., the -1st order mode, is usually the most obvious one, so only  $n = -1$  is considered in this paper. The scattering grating lobe of monostatic RCS appears when the -1st order mode moves in parallel fashion in opposite directions of the incident wave. We have

$$\sin(\theta) = \frac{\lambda_0}{2P}, \quad (2)$$

where  $\lambda_0$  is the wavelength in vacuum. When  $\lambda_0 > 2P$ , the -1st order mode is evanescent and the grating lobe is not detectable for the radar.

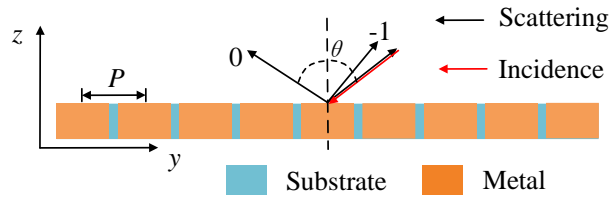


Fig. 1. Periodic structure illuminated by a plane wave.

The common phased array in the stealth platform is a periodic structure. We consider the array operating at 10 GHz as an example, as shown in Fig. 2. The linear array contains 8 elements, and its period is 15 mm. The substrate is Rogers 5880 with a relative dielectric constant of  $\epsilon_r = 2.2$  and a thickness of  $d = 3$  mm.

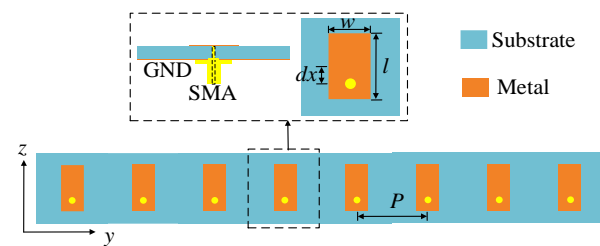


Fig. 2. Structure of the array with no metal wall, where  $dx = 2.4$  mm,  $w = 6.5$  mm, and  $l = 8.3$  mm.

When a plane wave at 12 GHz illuminates the array with an angle of  $\pm 56.4^\circ$ , the monostatic RCS at 12 GHz contains prominent scattering lobes, as shown in Fig. 3. The angle of the -1st scattering mode with  $P = 15$  mm is calculated from (2), as shown in Fig. 4.

To suppress the lobes, metal walls are placed around the array elements, as shown in Fig. 5. The metal walls

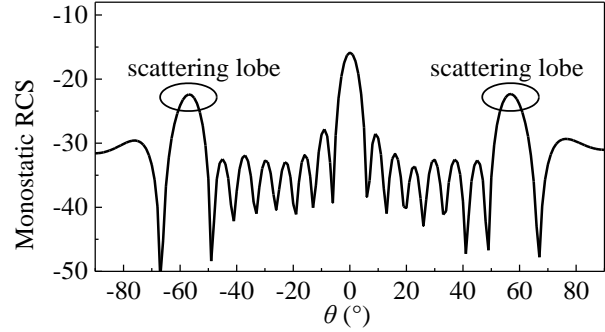


Fig. 3. Monostatic RCS of the array with no metal wall at 12 GHz.

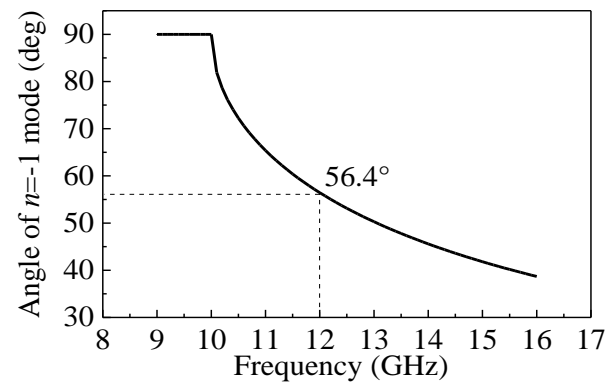


Fig. 4. Angle of the mode with the order of  $n = -1$ .

can reduce the scattering grating lobe and maintain the radiation performance of the array.

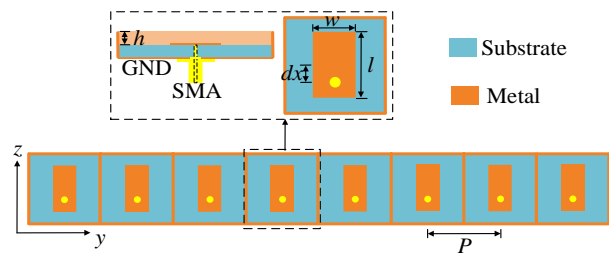


Fig. 5. Structures of the array with metal walls.

From [7], the incident light can be diffracted into the -1st order mode in a broad band by the metallic grating. When a designed metallic grating is placed around the antenna, the -1st order mode can be reduced. Here the metallic walls around the elements work as the metallic gratings. With well-designed metallic walls, the -1st order mode is reduced effectively.

### III. OPTIMIZATION DESIGN

The design of the array with metal walls requires numerous full-wave simulations, leading to a time-consuming process. The master-slave boundary and Floquet-Bloch port are used to carry out the element simulation instead of the array simulation. ANN with the Fourier series-based TF is utilized to accelerate the design procedure [5]. The input and output of ANN are the element geometric parameters of  $\mathbf{x} = [dx, h, l, w]^T$  and the EM response, respectively.

The Fourier series-based TF is written as

$$H(f) = \sum_{k=0}^{N_{order}} (a_k \cos(kq2\pi f) + b_k \cos(kq2\pi f)), \quad (3)$$

where  $N_{order}$  is the order of TF,  $a_k$  and  $b_k$  are the TF coefficients,  $q$  is the scale factor, and  $f$  is the frequency. When  $k = 0$ , there exists a constant term of  $a_0$ . The Fourier series-based TF is used to represent the wideband EM response. With the help of TF, the output of ANN is the coefficients of  $a_k$  and  $b_k$  instead of the EM response at multiple sampling frequency points. A trained ANN can predict  $a_k$  and  $b_k$  quickly, and then obtain the wideband EM response from (3). The learning efficiency and generalization of a wideband model can be improved with TF. The Fourier series-based TF only involves real variables and is suitable for real function fitting [5].

With the design of experiment (DOE) method [8], 49 training samples with seven-level and 16 testing samples with four-level defined in Table 1 are obtained from full-wave simulations. The EM response in this paper contains the magnitude of the -1st order Floquet-Bloch mode at 12 GHz when the incident angle is  $\pm 56.4^\circ$ , and the voltage standing wave ratio (VSWR) of the array is in the range of 9.5-10.5 GHz. Here, we trained two branch ANNs.

Table 1: Definition of training data and testing data for the array

| Geometric Parameters | Training Data (49) |      |       | Testing Data (16) |     |      |
|----------------------|--------------------|------|-------|-------------------|-----|------|
|                      | Min                | Max  | Step  | Min               | Max | Step |
| $dx$ (mm)            | 1.6                | 2.4  | 0.13  | 1.9               | 2.2 | 0.1  |
| $h$ (mm)             | 1.76               | 2.64 | 0.146 | 1.8               | 2.1 | 0.1  |
| $l$ (mm)             | 7.2                | 10.8 | 0.6   | 8.2               | 10  | 0.6  |
| $w$ (mm)             | 4.8                | 7.2  | 0.4   | 5.4               | 6.6 | 0.4  |

ANN<sub>1</sub>, with the Fourier series-based TF, is used to learn the relationship between the geometric parameters and VSWR. The order of the Fourier series-based TF in ANN<sub>1</sub> is chosen as 4 and then the total number of Fourier series coefficients is 9 [5]. From (3), the dimension of ANN<sub>1</sub> output parameters is 9.

ANN<sub>2</sub>, with no TF, is used to learn the relationship between the geometric parameters and the magnitude of

the -1st order Floquet-Bloch mode. The scattering grating lobe at 12 GHz only appears when the incident angle is equal to  $\pm 56.4^\circ$ . Because of the axial symmetry of the array, the magnitude of the scattering grating lobe at  $56.4^\circ$  is equal to that at  $-56.4^\circ$ . The output of ANN<sub>2</sub> represents the magnitude of the -1st order mode at the incident angle of  $56.4^\circ$ , so TF is not employed for ANN<sub>2</sub> due to the one-dimensional output. The training process of the two branch ANNs is shown in Fig. 6.

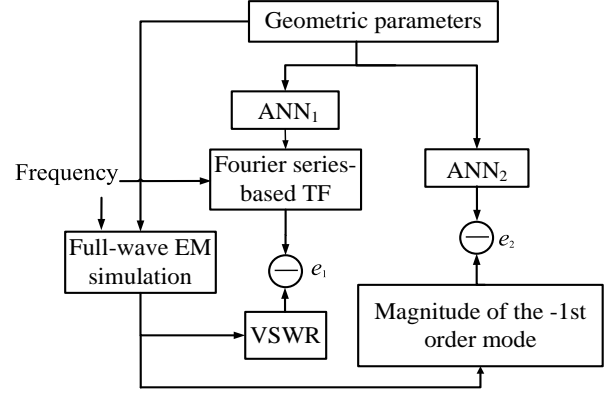


Fig. 6. Training process of the whole ANN model.

The node numbers of the hidden layers of ANN<sub>1</sub> and ANN<sub>2</sub> are determined by the Hecht–Nelson method [9]. When the node number of the input layer is  $n_{node}$ , the node number of the hidden layer is  $2n_{node} + 1$ . The root mean squared errors of  $e_1$  for ANN<sub>1</sub> and  $e_2$  for ANN<sub>2</sub> are provided below.

$$e_1 = \sqrt{\frac{\sum_{i=1}^{N_{sample}} (VSWR_i^{ANN} - VSWR_i^{simulation})^2}{N_{sample}}}, \quad (4)$$

$$e_2 = \sqrt{\frac{\sum_{i=1}^{N_{sample}} (mag_i^{ANN} - mag_i^{simulation})^2}{N_{sample}}}, \quad (5)$$

where  $VSWR_i^{simulation}$  is the VSWR calculated from the full-wave simulation of the  $i$ th ( $1 \leq i \leq N_{sample}$ ) sample,  $VSWR_i^{ANN}$  is the VSWR calculated from TF whose coefficients of  $a_k$  and  $b_k$  are obtained by the trained ANN<sub>1</sub>, and  $N_{sample}$  is the number of the samples.  $mag_i^{simulation}$  is the magnitude of the -1st order Floquet-Bloch mode calculated from the full-wave simulation of the  $i$ th sample, and  $mag_i^{ANN}$  is obtained from the trained ANN<sub>2</sub>.

After the training process, an independent dataset that is never used in training is applied to the testing process. The training and testing errors of ANN<sub>1</sub> are 1.78% and 1.52%, respectively, and the training and testing errors of ANN<sub>2</sub> are 1.04% and 0.93%, respectively. The errors of the ANNs are acceptable. Table 2 summarizes the brief information of the two ANN models.

Table 2: Brief information of two branch ANNs for array modeling

|                | ANN <sub>1</sub>                       | ANN <sub>2</sub>             |
|----------------|--|------------------------------|
| Input          | $\mathbf{x}$                           | $\mathbf{x}$                 |
| Output         | Coefficients of TF ( $a_k$ and $b_k$ ) | Magnitude of the $n=-1$ mode |
| ANN structure  | 4-9-9                                  | 4-9-1                        |
| TF             | Yes                                    | No                           |
| Training error | 1.78%                                  | 1.04%                        |
| Testing error  | 1.52%                                  | 0.93%                        |

After the two ANNs are well trained, the genetic algorithm (GA) is utilized to design the antenna array. GA repeatedly calls the trained ANNs until the optimization objective is achieved, where the optimization objective is the passband of 9.8-10.2 GHz with VSWR < 2 and the magnitude of -1st order mode is small enough. The optimized geometric parameters are  $\mathbf{x}_{opt} = [2.4, 2.5, 8.3, 6.5]^T$ . The monostatic RCS of the array, calculated by the full-wave simulation with  $\mathbf{x}_{opt}$ , at 12 GHz is shown in Fig. 7. From the figure, the grating lobe is suppressed by 5.8 dB, and the monostatic RCS at the angle of  $56.4^\circ$  is significantly reduced.

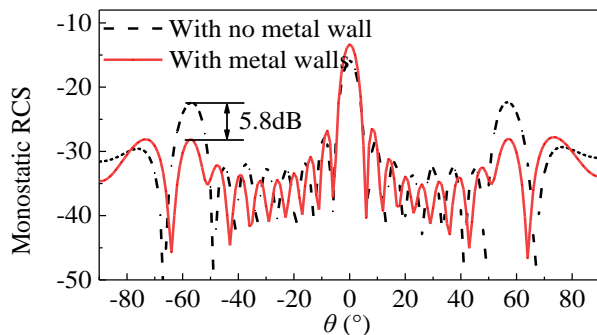


Fig. 7. Monostatic RCS of the array at 12 GHz.

The radiation performance is also considered here. The VSWRs of the optimized arrays with metal walls and without metal wall are provided in Fig. 8. The bandwidth of the array with metal walls is a little narrower than that of the array with no metal wall. Radiation patterns of the array at 10 GHz in the  $xoy$ -plane are also provided, as shown in Figs. 9 and 10. From Figs. 9 and 10, the gains of the arrays are reduced by 0.9 dB in the normal direction and improved by 1 dB at  $-50^\circ$ . The metal wall works as a wide-angle scanning structure, and the change of the radiation performance is acceptable.

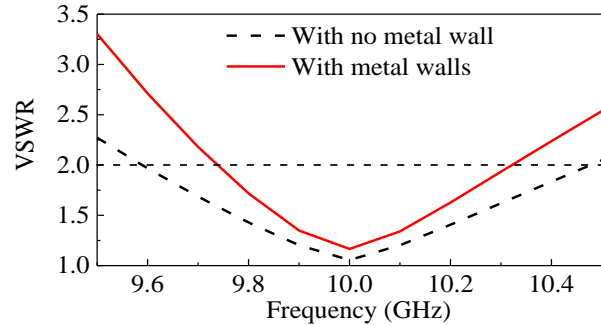
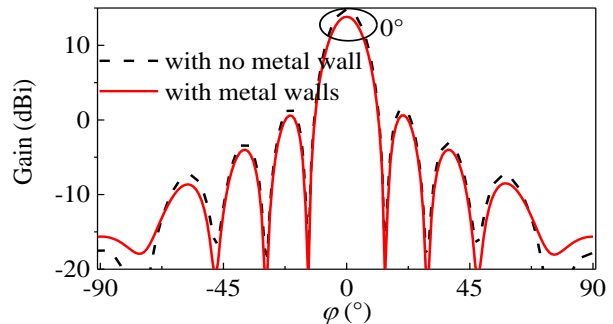
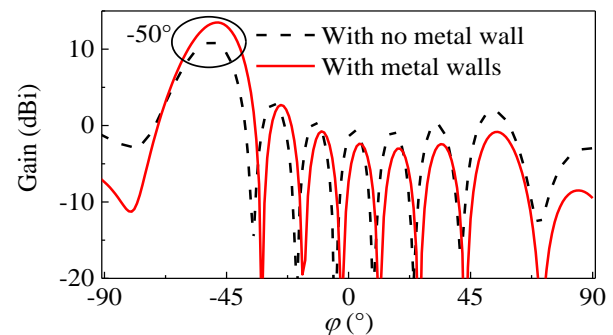


Fig. 8. Simulated VSWR of the center elements.

Fig. 9. Radiation patterns of arrays at 10 GHz in the  $xoy$ -plane with  $\varphi = 0^\circ$ .Fig. 10. Radiation patterns of arrays at 10 GHz in the  $xoy$ -plane with  $\varphi = -50^\circ$ .

#### IV. CONCLUSION

In this paper, an array with low scattering grating lobes is proposed. By placing well-designed metallic walls around the array elements, the grating lobe is suppressed and the change of the radiation performance is acceptable. ANN with the Fourier series-based TF, as a substitute for the full-wave simulation, is introduced to accelerate the optimization procedure. The proposed method can also be extended to the application

of two-dimensional (2-D) arrays, and we will study 2-D planar arrays with good radiation/scattering performance in future work.

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