Method of Moments Enhancement Technique for the Analysis of Sierpinski Pre-Fractal Antennas

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Abstract—The numerical analysis of highly iterated Sierpinski microstrip patch antennas by method of moments (MoM) involves many tiny subdomain basis functions, resulting in a very large number of unknowns. The Sierpinski pre-fractal can be defined by an iterated function system (IFS). As a consequence, the geometry has a multilevel structure with many equal subdomains. This property, together with a multilevel matrix decomposition algorithm (MLMDA) implementation in which the MLMDA blocks are equal to the IFS generating shape, is used to reduce the computational cost of the frequency analysis of a Sierpinski based structure.

Index Terms—Antennas, fractals, integral equation, method of moments (MoM).

I. INTRODUCTION

TRUCTURES based on the Sierpinski fractal are particularly interesting due to their multiband behavior [1], [2]. The radiation parameters of the antenna can be numerically computed using integral equation methods discretized by method of moments (MoM) [3].

In the case of highly iterated pre-fractal structures, there are many small geometry details that require tiny MoM subdomain basis functions for an accurate discretization of the induced current. This, together with the fact that the multiband antenna is electrically large at the highest operating bands, leads to a very large number of unknows (N) in the MoM formulation. The computational requirements to solve the full linear system using conventional methods (memory increases as N^2 and CPU time as N^3) can easily overcome the capabilities of desktop computer systems.

This paper will tackle the optimization of the MoM solution taking advantage of the geometrical properties of the iterated function system (IFS) [10] that generates the antenna geometry. Since the IFS is inherently multilevel, the most suitable MoM acceleration algorithms here are multilevel domain subdivision methods that can use Rao, Wilton, and Glisson linear triangle basis functions (RWG) [9], namely the multilevel fast

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multipole algorithm (MLFMA) [6] and the multilevel matrix decomposition algorithm (MLMDA) [7], [8]. This paper will present results only for the MLMDA, but the optimization strategies introduced here can be easily implemented also in the MLFMA.

II. IFS

Like most pre-fractal structures, the Sierpinski antenna can be built by using the concept of IFS [10]. Every IFS iteration is defined by a set of Q affine transformations in the plane $\{\omega_a\}_{a=1}^Q \colon R^2 \to R^2$ which can be written as

$$\omega_{q}(x) = A_{q}x + t_{q}$$

$$= \begin{pmatrix} r_{q1}\cos\theta_{q1} & -r_{q2}\sin\theta_{q2} \\ r_{q1}\sin\theta_{q1} & r_{q2}\cos\theta_{q2} \end{pmatrix} \begin{pmatrix} x_{1} \\ x_{2} \end{pmatrix}$$

$$+ \begin{pmatrix} t_{q1} \\ t_{q2} \end{pmatrix}$$

$$(1)$$

where x_1 and x_2 are the coordinates of point x. If x_1 and x_2 are the coordinates of point x. If $r_{q1} = r_{q2} = r_q$, with $0 < r_q < 1$, and $\theta_{q1} = \theta_{q2} = \theta_q$, the IFS transformation is a contractive similarity (angles are preserved) where r_q is the scale factor and θ_q is the rotation angle. The column matrix t_q is just a translation on the plane. Fig. 1 shows the Sierpinski fractal obtained from a single triangle after applying a set of transformations recursively.

Structures generated by an IFS necessarily have many equal subdomains at different levels. The impedance matrix resulting from MoM discretization has therefore plenty of redundant information, since the interaction between equal pairs of subdomains produces equal submatrices (1A, 2A, 3A in Fig. 2) if the Green's function has translation symmetry, as is the case here. There are many sets of equal submatrices in [Z] at different levels. The MoM implementation presented here uses the IFS definition in order to avoid recomputation and storage of redundant matrix elements.

III. MULTILEVEL ALGORITHMS

The electric field integral equation (EFIE) in the frequency domain discretized by MoM may be expressed in matrix form as [3], [9]

$$-[E_i] = [Z][J] \tag{2}$$

where [J] are the coefficients of the induced current discretized in RWG basis functions (unknowns), $[E_i]$ is the discretization of the incident field and [Z] is the impedance matrix. This matrix includes the Green's function with all the information about the multilayer media.

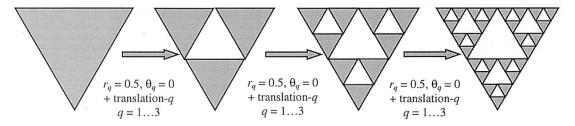


Fig. 1. Four iteration Sierpinski fractal obtained after a set of affine transformations.

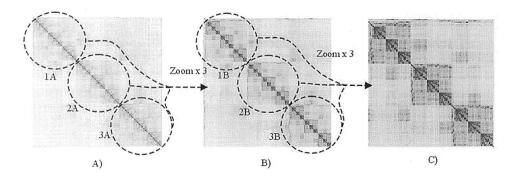


Fig. 2. Method of moments impedance matrix for a five iteration Sierpinski patch antenna. Grey scale shows the magnitude of the matrix elements. If the structure is generated by an IFS, the impedance matrix has plenty of redundant information: submatrices 1A, 2A and 3A are identical. Besides, due to the recursive application of the IFS, each one of these submatrices has also redundant information, as shown in B and C.

The induced current coefficients [J] are found using the Generalized Minimum Residual (GMRES) [4] iterative method. In each iteration, the main computational effort to obtain the kth estimation of the induced current $[J^{(k)}]$ are the matrix-vector products $[Z][J^{k-1}]$. Using direct matrix-vector multiplication, the operation count and the memory requirements for each iteration are proportional to N^2 .

A. Multilevel Subdivision of the Object

In order to reduce the operation count in the direct matrix-vector multiplication from N^2 to $N \log N$, the MLMDA and the MLFMA divide the object into an octal tree in three dimensions or a quad tree in two dimensions containing many nonoverlaping subdomains or boxes. The quad tree domain subdivision that is generally used for arbitrary structures is applied here to a Sierpkinski fractal in Fig. 3.

The interaction between a pair of subdomains can be computed as

$$[E_m] = [Z_{mn}][J_n] \tag{3}$$

where n and m, respectively, are the indexes of the RWG basis functions in the source and observation subdomains and $[Z_{mn}]$ is a submatrix of the impedance matrix. If all possible pairs of subdomains are considered, the matrix vector multiplication $[Z][J^{(k-1)}]$ may be obtained as the addition of submatrix operations of the form (3).

B. MLMDA

The MLMDA will be only outlined here, more details can be found in [8]. It consists of a recursive procedure that begins at level 2 and stops at the finest level L. For each nonempty source and observation boxes which belong to the same subdivision level $l(2 \le l \le L)$ there are two possible cases:

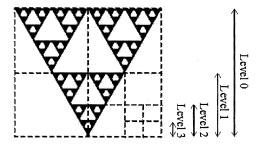


Fig. 3. Multilevel decomposition of the box enclosing a Sierpinski fractal in square subdomains.

- 1) Boxes are touching one another or are the same: then they are subdivided into level l+1 boxes, except if we have already reached the finest level, l=L. If this is the case, direct submatrix-vector multiplication (3) is performed, requiring the previous computation of the corresponding matrix terms of $[Z_{mn}]$. Time and memory requirements are proportional to M_bN_b where M_b and N_b are the number of original RWG functions in the source and observation box, respectively.
- 2) Boxes are not touching each other: In this case, the number of degrees of freedom (DoF) of the electric field $[E_m]$ at the observation box in (3) is smaller than N_b , and decreases for larger box-to-box distance and for smaller box size [7], [8]. Therefore, $[E_m]$ in (3) can be computed very efficiently in much less than M_bN_b operations by replacing the original currents by a very small set of equivalent ones that radiate the same field.

The computational cost of this recursive algorithm increases with the number of nonempty boxes in the multilevel subdivision of a given object. If we realize that boxes do not need to be square, as in Fig. 3, while they do not overlap and cover the whole geometry of the antenna, we can use boxes of the same

TABLE I COMPUTATIONAL REQUIREMENTS FOR THE CONVENTIONAL MOM SOLUTION USING DIRECT INVERSION ($f=9~{\rm GHz}$): Memory Grows as N^2 and Time as N^3 . There is Not Enough Computer Memory to Store the Impedance Matrix of the Seven Iteration Configuration

	Building	[Z]	Direct inversion	Total	Total	
	Mem. (MB)	t (s)	t (s)	Mem. (MB)	Time (s)	
4 it., N=568	4.9	3	1	4.9	4	
5 it., N=1702	44.2	21	14	44.2	35	
6 it., N=5104	397.5	176	383	397.5	559	
7 it., N=15310	3577.5			3577.5		

TABLE II Computational Requirements for the MoM Solution Using the GMRES Iterative Method $(f=9~\mathrm{GHz})$. Time Performance Has Improved, However the Seven Iteration Configuration Cannot be Analyzed Yet

	Building [Z]		Preconditioning		10.11	Iteratio	n	Total	Total
	Mem(MB)	t (s)	Mem(MB)	t(s)	n. it	t per it(s)	It.time (s)	Mem.(MB)	Time (s)
4 it., N=568	4.9	3	0.4	0.2	5	0.3	1.5	5.3	5
5 it., N=1702	44.2	21	3.8	2	4	0.6	2.4	48.0	25
6 it., N=5104	397.5	176	27.3	25	3	6.0	18	424.8	219
7 it, N=15310	3577.5								

TABLE III COMPUTATIONAL REQUIREMENTS FOR THE MOM SOLUTION USING GMRES AND THE MLMDA SCHEME WITH SQUARE BOXES ($f=9~\mathrm{GHz}$). Time and Memory Requirements are Smaller Than the Two Previous Strategies, Specially When N Grows

	Building [Z]		Preconditioning			Iteratio	n	Total	Total
	Mem(MB)	t (s)	Mem(MB)	t (s)	n. it	t per it(s)	It.time (s)	Mem.(MB)	Time (s)
4 it., N=568	3.7	3	0.4	0.2	5	0.1	0.5	4.1	3.7
5 it., N=1702	25.3	12	3.8	1.5	4	0.3	1.2	29.1	15
6 it., N=5104	102.1	50	27.3	17	3	1.5	4.5	129.4	71
7 it, N=15310	375.1	184	128.3	136	9	8	72	503.4	392

shape and size as the IFS building blocks. For the case of the Sierpinski antenna the optimum choice is triangular boxes. This results in a multilevel subdivision with: a) many empty boxes and b) many pairs of source and field boxes having the same interaction matrices either in step 1 or in the equivalent source formulation of step 2. These matrices will be computed once, stored in memory and reused whenever required, leading to a dramatic reduction in computation time and memory storage.

IV. RESULTS

Configurations of a microstrip Sierpinski patch antenna with four, five, six, and seven iterations have been used as a benchmark in order to test different approaches to enhance the MoM analysis. In all the cases the scale factor is 2 and the height of the equilateral triangle defining the Sierpinski patch (level 0 in Fig. 3) is 8.89 cm. The dielectric substrate is 1.57 mm thick with a relative permitivity of 2.33. The patch was excited with a standard coaxial probe located in the lower corner of the Sierpinski fractal.

The computer used in the simulation is a desktop PC with an AMD Athlon CPU at 1.33 GHz and 1.5 GB of RAM. The programming language is MATLAB 6 with time-critical routines coded in C language.

Table I shows the computational requirements for the solution of the MoM linear system (2) using direct matrix inversion. Memory requirements grow as N^2 and time as N^3 , as expected. For the seven iteration configuration, the storage requirements overcome the available memory.

In Table II the same test is repeated using GMRES instead of the direct inversion. Preconditioning is used to reduce the number of iterations [4]. The preconditioner is an sparse matrix that includes all the impedance matrix elements corresponding to basis and testing functions that are close to each other, and zeros elsewhere. The incomplete LU decomposition [4] allows fast multiplication by the inverse of the preconditioner at each iteration. The extra time used in the preconditioning and the small extra increase in memory are more than compensated by the very small number of iterations. The time for building the matrix remains the same as before, while frequently the GMRES iterative solution is one order of magnitude faster. It is therefore the matrix filling time which now becomes the bottleneck.

The MLMDA scheme can be used to compute the matrix vector multiplication $[Z][J^{(k-1)}]$ in GMRES, instead of building the whole [Z] matrix. Table III shows the computational requirements of MLMDA with the commonly used square box subdivision (Fig. 3). The matrix filling time and the iteration time are greatly reduced, since MLMDA does not need to compute explicitly all the elements of [Z]. This reduction is particularly important for the large problem of 15 310 unknowns, in which the MoM [Z] matrix does not fit in computer memory.

The last level of optimization is reached when triangular boxes are used in the MLMDA approach and IFS symmetries are exploited in order to avoid recomputation of redundant MLMDA matrices (Table IV). Memory requirements increase only linearly with the number of unknowns, while the CPU time per iteration increases roughly as $N \log N$. Matrix filling is no more the slowest part of the process. This is due to the

TABLE IV

COMPUTATIONAL REQUIREMENTS FOR THE MOM SOLUTION USING GMRES AND THE MLMDA SCHEME WITH TRIANGULAR BOXES $(f=9~\mathrm{GHz})$. Taking Advantage of Redundancies and the Multilevel Structure of the Object, the Memory Requirements are Very Small and Grow as N While the Time Per Iteration Increases Roughly as $N\log N$. Optimization of Preconditioning is the Open Point

	Building [Z]		Preconditioning			Iterati	on	Total	Total
	Mem(MB)	t (s)	Mem(MB)	t(s)	n. it	t×it(s)	It.time (s)	Mem.(MB)	Time (s)
4 it., N=568	1.4	0.8	0.4	0.2	5	0.03	0.15	1.8	1.1
5 it., N=1702	4.1	2	3.8	1.5	4	0.17	0.68	7.9	4.1
6 it., N=5104	12.0	7	27.3	16	3	0.65	1.95	39.3	25
7 it, N=15310	38.2	19	128.3	133	9	2.60	23.4	166.5	176

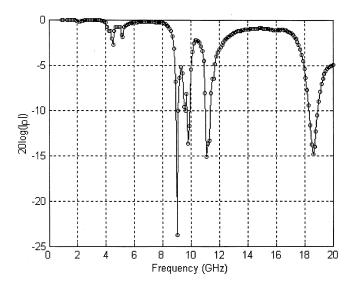


Fig. 4. Input reflection coefficient versus frequency for a seven iteration Sierpinski patch antenna.

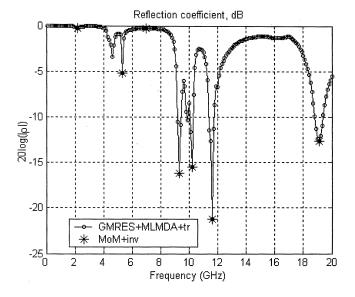


Fig. 5. Input reflection coefficient versus frequency for a six iteration Sierpinski patch antenna. It can be seen as there is not a noticeable loss of accuracy in the computation with GMRES + MLMDA with triangular boxes respect to the solution with direct inversion.

fact that it is only necessary to compute very few interactions between boxes.

The differences between the reference brute force approach of Table I and the optimized algorithm in Table IV are impressive, allowing us the analysis of this structure in a wide range of frequencies (Fig. 4). It must be pointed out that, for the problem of 15 310 unknowns, the preconditioning, which has not been optimized for fractal structures yet, is taking most of the computational effort.

Finally, Fig. 5 shows the reflection coefficient of a six iteration Sierpinski antenna. The whole curve (200 frequencies) obtained with GMRES + MLMDA with triangular boxes is calculated in about the same time than seven specific frequencies with the standard MoM, without any noticeable loss of accuracy.

V. CONCLUSION

Computational requirements in the analysis of IFS defined pre-fractal antennas by MoM can be much reduced by making use of the redundancy and the multilevel structure of this geometry.

It has been shown that, for the case of a Sierpinski pre-fractal antenna, the combination of the GMRES and the MLMDA scheme, together with the appropriate choice of the shape of the boxes in the multilevel subdivision, leads to a very efficient solution. Our best implementation (Table IV) produces a reduction by a factor of 20 in the total computation time and a factor of 10 in the total memory, compared with a direct application of MoM (Table I). The bottleneck in the MoM analysis of IFS defined geometries is now in the preconditioning, which has not been optimized for IFS structures yet.

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