Quantitative Evaluation for Computational Cost of CG-FMM on Typical Wiregrid Models

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1. Introduction

The Method of Moments (MoM) is one of the powerful techniques for numerical analysis of antennas and scatterers [1], [2]. However, CPU time and computer memory to solve matrix equation in MoM by using conventional direct methods such as Gaussian elimination are proportional to $O(N^3)$ and $O(N^2)$, respectively, where $N$ is the number of unknowns. Even though powerful computers with large memories are available, it is impossible to solve large-scale problems with millions of unknowns by the direct methods. Therefore, many techniques have been developed to reduce the CPU time and computer memory of MoM computation.

Iterative method such as the Conjugate Gradient (CG) method has been used to solve the MoM matrix equation in previous studies [3]–[8]. Unlike direct methods, the CPU time to solve the matrix equation by CG method is determined by number of iteration steps and CPU time per iteration. Therefore, computational cost of CG method is classified into three factors, i.e., the number of iteration steps, CPU time per iteration, and computer memory. The number of iteration steps depends on the analysis model, while the CPU time per iteration and computer memory of CG method are both the order of $O(N^2)$.\textsuperscript{1}

In above three factors, the number of iteration steps of CG method has been discussed in a few papers so far and it has been reported that the number of iteration steps is strongly depended on the condition number of the impedance matrix $Z$ [6]–[8]. But in these papers, little attention is paid to relation between the number of iteration steps and analysis model, as well as relation between condition number of $Z$ and analysis model.

On the other hand, it is known that other two factors except the number of iteration steps, namely, the CPU time per iteration and computer memory of CG method can be reduced by the Fast Multipole Method (FMM) [9], [10] or Multi-Level Fast Multipole Algorithm (MLFMA) [11]–[20]. It has been considered that the CPU time per iteration and computer memory of CG method combined with FMM (CG-FMM) are both $O(N^{1.5})$. However, the CPU time per iteration and computer memory in CG-FMM depend on many parameters for FMM as well as analysis model. Choice of parameters such as the number of multipoles or a group size for the two and three dimensional FMM have been discussed in a few papers [21], [22]. However, parameter choice based on these papers are not always correct in general since shape of analysis model, which is highly related to the number of multipoles $L$, has not been considered in these papers.

In this paper, four typical wiregrid models, which is formed electrically continuous or separated geometry, and one-dimensional or two-dimensional geometry, are introduced. It is shown that relation between the number of multipoles $L$ and number of segments in each group $K$ is determined by dimension of segment arrangement in an analysis model, and the relation is used to discuss not only the CPU time per iteration but also the computer memory required for analysis. Based on numerical simulation for these typical models, mutual relation among the number of iteration steps, condition number of $Z$ and analysis model can be discussed universally from a physical point of view.

This paper is organized as follows. Section 2 presents principle of CG-FMM and important parameters related to the computational cost. In Sect. 3, four typical models and two different relations between $L$ and $K$ are shown. In Sect. 4, the computational cost of CG-FMM for these typical models is discussed theoretically and numerically.

2. Principle of CG-FMM

2.1 Conjugate Gradient Method

Matrix equation produced by the MoM is expressed by

$$ZI = V,$$

where $Z$ is $N \times N$ impedance matrix produced by the MoM,
where \( t \) is \( N \)-dimensional known voltage vector, and \( I \) is \( N \)-dimensional unknown current vector. The algorithm of CG method for solving Eq. (1) is summarized as follows [3]–[8].

**CG method for MoM.**

- After initial value of \( I \) is set to be \( I_0 \), initial value of residual vector \( R_0 \) and correction vector for solution \( P_0 \) are calculated as
  \[
  R_0 = V - ZI_0,
  P_0 = Z' R_0.
  \]
- Iterations \( (i = 1, 2, \ldots) \) are stopped after the desired residual norm \( ||R_i|| \) is obtained as
  \[
  \alpha_i = \frac{\langle ZP_{i-1}, R_{i-1} \rangle}{||ZP_{i-1}||^2},
  \]
  \[
  I_i = I_{i-1} + \alpha_i P_{i-1},
  \]
  \[
  R_i = V - ZI_i = R_{i-1} - \alpha_i ZP_{i-1}.
  \]
  If \( ||R_i|| < \epsilon ||V|| \), iteration is stopped.
  \[
  \beta_i = \frac{||Z'R_i||^2}{||Z'R_{i-1}||^2},
  \]
  \[
  P_i = Z'R_i + \beta_i P_{i-1},
  \]

where \( \alpha_i \) and \( \beta_i \) are correction coefficients for \( I_{i-1} \) and \( P_{i-1} \), respectively, and \( \epsilon \) is an error control parameter for the solution. \( Z' \) denotes the conjugate transpose of \( Z \).

In the above algorithm, two times of matrix-vector multiplication are carried out in each iteration and CPU time for the matrix-vector multiplication is \( O(N^2) \). In addition, computer memory for storing \( Z \) is also \( O(N^2) \).

2.2 Fast Multipole Method

FMM is a method based on the addition theorem of the scalar Green’s function [9], [10]. FMM can reduce both the CPU time for the matrix-vector multiplication and computer memory of CG method to \( O(N^{1.5}) \) when number of groups \( M = \sqrt{N} \). The matrix-vector multiplication in each step of CG-FMM can be collectively carried out by grouping scheme based on the addition theorem. In addition, mutual impedance between far segments are calculated by the addition theorem at every time of the matrix-vector multiplication in CG-FMM and the full impedance matrix \( Z \) is not necessary to be stored. In what follows, formulation of the mutual impedance between far segments based on FMM is explained.

In Galerkin-MoM, the expression of the mutual impedance between far segments is obtained as

\[
Z_{mm'k'k}^{far} = j \omega \mu_0 \int_{l_{mk}} f_{mk}(r_k) \cdot \int_{l_{mk'}} f_{mk'}(r_{k'})dr' \cdot dr,
\]

where \( m \) and \( m' (= 1, 2, \ldots, M) \) are observation group number and source group number, respectively [1]. \( k \) and \( k' (= 1, 2, \ldots, N/M(=K)) \) represent segment number in \( m \)th observation group and \( m' \)th source group, respectively. Here, \( M \) is the total number of groups and \( K \) is the total number of segments in each group. \( l_{mk} \) and \( l_{mk'} \) denote the path of integration along the observation segment and source segment.

Figure 1 shows the definition of vectors and subscripts for FMM formulation.

\[
Z_{mm'k'k}^{far} = \omega \mu_0 \int_0^{2\pi} \int_0^\pi s_{mk}(\hat{k})T_{mm'}s_{mk'}(\hat{k}) \sin \theta d\theta d\phi,
\]

is obtained, where \( s_{mk}(\hat{k}) \), \( s_{mk'}(\hat{k}) \), and \( T_{mm'} \) are expressed as follows.

\[
s_{mk}(\hat{k}) = \int_{l_{mk}} e^{j k r_{mk}} (\hat{t} - \hat{k}) \cdot f_{mk}(r_k)dr_k.
\]

\[
s_{mk'}(\hat{k}) = \int_{l_{mk'}} e^{j k r_{mk'}} (\hat{t} - \hat{k}) \cdot f_{mk'}(r_{k'})dr_{k'}.
\]

\[
T_{mm'} = \sum_{l=0}^L (-j)^l \binom{2l+1}{l} h_l^{(2)}(k_0 r_{mm'})P_l(\hat{k} \cdot \hat{r}_{mm'}).
\]

\( s_{mk}(\hat{k}) \), called “receiving function” means the mutual impedance between the center of the \( m \)th observation group and \( k \)th segment in \( m \)th group. \( s_{mk'}(\hat{k}) \), called “radiation function” means the mutual impedance between the center of the \( m' \)th source group and \( k' \)th segment in \( m' \)th group. \( T_{mm'} \), called “transfer function” represents the mutual impedance between the center of the \( m \)th observation group and that of the \( m' \)th source group. \( \hat{t} = \hat{x} + \hat{y} + \hat{z} \) denotes the unit dyad and \( \hat{k} = \hat{x} \sin \theta \cos \phi + \hat{y} \sin \theta \sin \phi + \hat{z} \cos \theta \) is an unit vector of radial direction on an unit sphere. \( h_l^{(2)} \) is the \( l \)th order spherical Hankel function of the second kind and \( P_l \) is the Legendre polynomial of \( l \)th order.

Equation (6) is a finite series whose summation is truncated at \( L \). The truncation number \( L \) is given by a following empirical formula

\[
L = k_0 D_{max} + \alpha L \ln(k_0 D_{max} + \pi),
\]
where $D_{\text{max}}$ is the maximum diameter of the groups and $\alpha_L (= 0 \sim 10)$ is the error control parameter for the transfer function. In addition, $L$ is also used for the points of double numerical integration in Eq. (3). In order to obtain accurate results, $L$ point Gauss-Legendre integration in $\theta$ and $2L$ point trapezoidal integration in $\phi$ are applied to Eq. (3). Namely, each component of $s_m(\hat{k})$, $s_{m'k}(\hat{k})$ as well as $T_{mm'}$ are stored as $L \times 2L = 2L^2$ discrete data in numerical calculation.

According to above formulae, the matrix-vector multiplication for the mutual impedance elements between far segments in CG-FMM is represented by following three steps.

**Step 1** Aggregation step ($m' = 1, ..., M$)

$$S_m(\hat{k}) = \sum_{k'=1}^{K} s_{m'k}(\hat{k}) a_{k'}. \tag{8}$$

In $m'$th source group, the radiation function $s_{m'k}(\hat{k})$ is multiplied by $a_{k'}$, which is $k'$th component of a vector $a$, and results of the multiplication are stored. Here, $a$ is $I$, $R$, or $P$ (see Sect. 2.1) in CG-FMM. Since the aggregation step is applied to all $N$ radiation function, the CPU time for the aggregation step is $O(N \times 2L^2)$.

**Step 2** Translation step ($m = 1, ..., M$)

$$S_m(\hat{k}) = \sum_{m' = 1}^{M} \sum_{m' \in M_{\text{tar}}} S_{m'}(\hat{k}) T_{mm'}(k_0 r_{mm'}, \hat{k} \cdot \hat{r}_{mm'}). \tag{9}$$

$S_{m'}(\hat{k})$ is transferred to the center of $m'$th observation group by the transfer function $T_{mm'}$. $M_{\text{tar}}$ denotes all far groups from $m$th observation group. This translation step from $M$ source groups is repeated for all $M$ observation groups and the total CPU time for the translation step is $O(M^2 \times 2L^2)$.

**Step 3** Disaggregation step ($i = 1, ..., N$)

$$\sum_{j=0}^{N_{\text{tar}}} Z_{ij}^f a_j \approx \frac{\omega (d_x d_y)}{(4\pi)^2} \sum_{l=1}^{L} \sum_{\theta=1}^{2L} W(\theta) \frac{\pi}{L} \left[ s_m(\hat{k}) \cdot S_m(\hat{k}) \right]. \tag{10}$$

The multiplication for $Z_{ij}^f$ in the $i$th row of the impedance matrix is collectively calculated by right-hand side. Here, $N_{\text{tar}}$ denotes all far segments from segments in the $n$mth observation group, and $W(\theta)$ is the weight of Gauss-Legendre integration in $\theta$. The disaggregation step is carried out for all $N$ rows of the impedance matrix and the CPU time for the disaggregation step is $O(N \times 2L^2)$.

The CPU time for all steps in the multiplication of $Z_{ij}^f$, which denotes the mutual impedance between near segments, and that of $Z_{ij}^s$ are shown in Table 1. The computer memory for CG-FMM is also tabulated in Table 2.

### 3. Typical Models

Four typical models described in the paper are shown in Fig. 2. Figure 2(a) shows one-dimensional side-by-side dipole array antenna (Model 1), where dipole element, whose length is $l$ and radius is $a$, is arrayed periodically. Each element is divided into $K$ wire segments. Periodicity of the group and segment location are both satisfied in this model.

Figure 2(b) shows long dipole antenna (Model 2), whose length is $l$ ($\gg a$) and radius is $a$. The dipole antenna is divided into $N$ wire segments. Periodicity of the group and segment location are both satisfied in this model.

Figure 2(c) shows two-dimensional planar array of conducting scatterers (Model 3). A planar element, whose size is $d_x \times d_y$, is arrayed periodically. Each element is divided into $K$ wire segments with radius $a$. Periodicity of the group and segment location are both satisfied in this model.

Figure 2(d) shows planar conductor (Model 4). A planar element, whose size is $d_x \times d_y$, is divided into $N$ wire segments with radius $a$. Periodicity of the group location is only satisfied in this model. Since the uniform grouping for segments is difficult, non-uniform grouping scheme and versatile program using CG-FMM is used for analysis of the model.

Relation between the number of multipole $L$ and segments in each group $K$ is shown in Table 3. In general, $K$ is proportional to $D_{\text{max}}^2$ for two-dimensional arrangement, while $K$ is proportional to $D_{\text{max}}$ for one-dimensional arrangement. Substituting the relation between $K$ and $D_{\text{max}}$ into Eq. (7), the relation between $L$ and $K$ shown in Table 3 can be easily derived.

As shown in Table 4, the use of periodicity in CG-FMM can reduce the computer memory required for analysis. By using Tables 2–4, the computer memory required for CG-FMM corresponding to each analysis model can be easily obtained. In a similar way, CPU time per iteration in CG-FMM can be also obtained from Tables 1 and 3.

<table>
<thead>
<tr>
<th>Step</th>
<th>Type of multiplication</th>
<th>CPU time</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Multiplication for $Z_{ij}^f$</td>
<td>$O(MK^2)$</td>
</tr>
<tr>
<td>2</td>
<td>Multiplication for $Z_{ij}^s$</td>
<td>$O(2NL^2)$</td>
</tr>
<tr>
<td>3</td>
<td>Multiplication for $Z_{ij}^f$</td>
<td>$O(2ML^2)$</td>
</tr>
</tbody>
</table>

### Table 1: CPU time for matrix-vector multiplication in CG-FMM.

<table>
<thead>
<tr>
<th>Type of multiplication</th>
<th>Step</th>
<th>CPU time</th>
</tr>
</thead>
<tbody>
<tr>
<td>Multiplication for $Z_{ij}^f$</td>
<td>Step 1</td>
<td>$O(2NL^2)$</td>
</tr>
<tr>
<td>Multiplication for $Z_{ij}^s$</td>
<td>Step 2</td>
<td>$O(2ML^2)$</td>
</tr>
<tr>
<td>Multiplication for $Z_{ij}^f$</td>
<td>Step 3</td>
<td>$O(2NL^2)$</td>
</tr>
</tbody>
</table>

### Table 2: Computer memory for CG-FMM.

<table>
<thead>
<tr>
<th>Stored contents</th>
<th>Computer memory</th>
</tr>
</thead>
<tbody>
<tr>
<td>$V_i, I, P, R$</td>
<td>$O(N)$</td>
</tr>
<tr>
<td>$W(\theta)$</td>
<td>$O(L)$</td>
</tr>
<tr>
<td>$Z_{ij}^f$</td>
<td>$O(MK^2)$</td>
</tr>
<tr>
<td>$s_m(\hat{k})$</td>
<td>$O(2NL^2)$</td>
</tr>
<tr>
<td>$T_{mm'}(k_0 r_{mm'}, \hat{k} \cdot \hat{r}_{mm'})$</td>
<td>$O(2ML^2)$</td>
</tr>
</tbody>
</table>

### Table 3: Computer memory for CG-FMM.

<table>
<thead>
<tr>
<th>Stored contents</th>
<th>Computer memory</th>
</tr>
</thead>
<tbody>
<tr>
<td>$V_i, I, P, R$</td>
<td>$O(N)$</td>
</tr>
<tr>
<td>$W(\theta)$</td>
<td>$O(L)$</td>
</tr>
<tr>
<td>$Z_{ij}^f$</td>
<td>$O(MK^2)$</td>
</tr>
<tr>
<td>$s_m(\hat{k})$</td>
<td>$O(2NL^2)$</td>
</tr>
<tr>
<td>$T_{mm'}(k_0 r_{mm'}, \hat{k} \cdot \hat{r}_{mm'})$</td>
<td>$O(2ML^2)$</td>
</tr>
</tbody>
</table>
4. Numerical Results

In this section, numerical analysis of four typical models is carried out using the Richmond’s MoM [2]. Dell Precision PWS 380 with 2 GB RAM is used for all numerical calculation. \( \varepsilon = 10^{-4} \) is used for the convergence criterion of CG-FMM and \( \alpha_L = 2 \) is employed to Eq. (7) for sufficient truncation number. In addition, the self-impedance and the mutual impedance between segments in adjacent groups are calculated by the MoM, rather than by FMM. Initial value of current vector \( \mathbf{I} \) is zero vector. Preconditioning technique is not used in both CG method and CG-FMM.

4.1 Number of Iteration Steps and Condition Number

The number of iteration steps required for CG method and CG-FMM and the condition number \( \kappa \) of \( \mathbf{Z} \) for four typical models are shown in Fig. 3. The condition number \( \kappa \) of \( \mathbf{Z} \) is defined by

\[
\kappa = \sqrt{\frac{\lambda_{\text{max}}}{\lambda_{\text{min}}}}.
\]

where \( \lambda_{\text{max}} \) and \( \lambda_{\text{min}} \) are the maximum and minimum eigenvalue of \( \mathbf{Z}^\top \mathbf{Z} \), respectively. As shown in Fig. 3, the tendency of the condition number \( \kappa \) is almost the same to that of the number of iteration steps determined by error control parameter. From those results, it is found that the condition
number is highly related to that of the number of iteration steps.

Large $\kappa$ means that the impedance matrix $Z$ is ill-conditioned and convergence of the solution obtained in iterative procedure is slow. On the other hand, small $\kappa$ means that the impedance matrix $Z$ is well-conditioned and convergence of the solution is fast.

From a physical point of view, condition number $\kappa$ indicates how the solution $I$ in Eq. (1) is sensitive to the values of the elements in the impedance matrix $Z$. When electrical connection between segments in the model is strong due to linear-connection of segments or strong mutual coupling such as Models 2 and 3, the condition number $\kappa$ increases as shown in Figs. 3(b) and 3(c), and $I$ is sensitive to $Z$. On the other hand, when electrical connection between segments in the model is weak due to small mutual coupling or grid-connection of segments such as Models 1 and 4, condition number $\kappa$ is small as shown in Figs. 3(a) and 3(d), and $I$ is not sensitive to $Z$. Constant $\kappa$ shown in Figs. 3(a) and 3(d) means that mutual coupling effect to an element does not change anymore even if $M$ increases. In general, since current of each segment is affected by all segments, the number of iteration steps required for analysis increases when number of segments $N$ increases. However, mutual coupling effect to $I$ from significantly far elements is very small and negligible in the array antenna such as Models 1 and 3. Therefore, $\kappa$ and number of iteration steps required for analysis of the array antenna which has enough number of elements becomes constant.

### 4.2 CPU Time per Iteration

The CPU time per iteration for analysis of four typical models is shown in Fig. 4. It is found that the CPU time per iteration of CG-FMM is $O(N^2)$ for the Models 1 and 2, while that for the Models 3 and 4 is reduced to $O(N^{1.5})$. Using $L \propto K$ shown in Table 3, it can be derived from Table 1 that the CPU time per iteration for analysis of one-dimensional models is $O(N^2)$, which is independent of the value of $M$, because the CPU time of Step 2 in Table 1 is $O(N^2)$. In Fig. 4(c), the order of the CPU time per iteration is $O(N)$, when $N$ is small. Since $K$ is constant (= 84) in the model 3, it is easy to understand that $K \gg M$ when $N(=MK)$ is as small as about 1000 or less. Based on $L^2 \propto K$ shown in
Table 5  Computational cost of CG-FMM for four typical wiregrid models.

<table>
<thead>
<tr>
<th>Item</th>
<th>Model 1</th>
<th>Model 2</th>
<th>Model 3</th>
<th>Model 4</th>
</tr>
</thead>
<tbody>
<tr>
<td>Properties of analysis model</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Segment arrangement</td>
<td>1D</td>
<td>1D</td>
<td>2D</td>
<td>2D</td>
</tr>
<tr>
<td>Segment connection</td>
<td>Linearly connected</td>
<td>Linearly connected</td>
<td>Grid connected</td>
<td>Grid connected</td>
</tr>
<tr>
<td>Number of elements</td>
<td>Array</td>
<td>Single</td>
<td>Array</td>
<td>Single</td>
</tr>
<tr>
<td>Type of periodicity used in CG-FMM</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Group location</td>
<td>Used</td>
<td>Used</td>
<td>Used</td>
<td>Used</td>
</tr>
<tr>
<td>Segment location</td>
<td>Used</td>
<td>Used</td>
<td>Used</td>
<td>Not used</td>
</tr>
<tr>
<td>Relation between L and K</td>
<td>( L \propto K )</td>
<td>( L \propto K )</td>
<td>( L^2 \propto K )</td>
<td>( L^2 \propto K )</td>
</tr>
<tr>
<td>Parameter setting on grouping</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>( K ) = Const.</td>
<td>( M = K = \sqrt{N} )</td>
<td>( K ) = Const.</td>
<td>( M = K = \sqrt{N} )</td>
<td>( M = K = \sqrt{N} )</td>
</tr>
<tr>
<td>Computational cost</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Number of iteration steps</td>
<td>Const.</td>
<td>( O(N) )</td>
<td>( O(N^{0.5}) \to Const. )</td>
<td>( O(N^{0.25}) )</td>
</tr>
<tr>
<td>CPU time per iteration</td>
<td>( O(N^2) )</td>
<td>( O(N^2) )</td>
<td>( O(N) \to O(N^{1.5}) )</td>
<td>( O(N^{1.5}) )</td>
</tr>
<tr>
<td>Total CPU time</td>
<td>( O(N^2) )</td>
<td>( O(N^3) )</td>
<td>( O(N^{1.5}) )</td>
<td>( O(N^{1.5}) )</td>
</tr>
<tr>
<td>Computer memory</td>
<td>( O(K^2) \to O(N) )</td>
<td>( O(N^{1.5}) )</td>
<td>( O(K^2) \to O(N) )</td>
<td>( O(N^{1.5}) )</td>
</tr>
</tbody>
</table>

Table 3 and \( K(=\text{Const.}) \gg M \), it can be obtained from Table 1 that the CPU time per iteration for analysis of model 3 is \( O(N) \) when \( N \) is small. From these results, it is concluded that the CPU time per iteration to analyze an antenna which has one-dimensional segment arrangement cannot be reduced by CG-FMM.

4.3 Computer Memory

The order of the computer memory required for analysis of each model can be theoretically derived from Tables 2–4 without running programs and is tabulated in Table 5. From
the Table 5, it is found that the computer memory required for analysis of any models is smaller than $O(N^2)$ in CG-FMM. For the array antenna/scatterer such as the Models 1 and 3, the computer memory for $S_{md}$ and $S_{me}$ is dominant for the total computer memory when $M$ is very small. On the other hand, the computer memory for $T_{mn}$ is dominant for the total computer memory when $M$ is larger than $K$. Therefore, the order of the computer memory required for analysis varies as $O(K^3) \rightarrow O(N)$ for Model 1 and $O(K^2) \rightarrow O(N)$ for Model 3.

The order of the number of iteration steps, CPU time per iteration, total CPU time and computer memory is tabulated in Table 5.

5. Conclusion

In this paper, relation between the computational cost of CG-FMM and the analysis model was quantitatively evaluated by numerical simulation for four typical wire antennas or wiregrid models. It is found that the number of iteration steps required for analysis by CG-FMM depends on the condition number of $Z$. From discussion of relation between the condition number $\kappa$ and electrical properties of the models, it is shown that the condition number $\kappa$ becomes large due to linear-connection of segments or strong mutual coupling between elements. Similarly, it is shown that the condition number $\kappa$ becomes small due to grid-connection of segments exist or weak mutual coupling between elements. The computer memory required for analysis are reduced from $O(N^2)$ for all models by the use of CG-FMM. On the other hand, it is shown that the CPU time per iteration can be reduced only for models which has two-dimensional segment arrangement in each group.

Since the models for analysis used in this paper are simple linear or planar models, the computational cost of CG-FMM for more complicated structures is desired to be evaluated. In addition, the computational cost of CG-FMM applied to the Rao-Wilton-Glisson (RWG) basis function [24] as well as the Schaubert-Wilton-Glisson (SWG) basis function [25] is expected to be evaluated for analysis of structures including dielectric bodies.

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References


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