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# AMG Smoothers for Maxwell's Equations

L. Claus, R. D. Falgout, M. Bolten

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# AMG SMOOTHERS FOR MAXWELL'S EQUATIONS \*

L. CLAUS †

In collaboration with: R. Falgout, M. Bolten,

**Abstract.** Algebraic Multigrid (AMG) is used to speed up linear system solves in a wide variety of applications. This paper concentrates on expanding AMG's applicability to important new classes of problems through algorithms that automatically construct advanced smoothing techniques when needed. In particular, we apply AMG to solve Maxwell's equations. These AMG relaxation methods have their roots in smoothing methods used for geometric multigrid methods. Arnold, Falk and Winther developed an overlapping Schwarz smoother for geometric multigrid and Hiptmair developed a distributive relaxation approach. We use this knowledge to construct new smoothing procedures for AMG. We develop adapted overlapping Schwarz smoothers and distributive relaxation for AMG. We use Nédélec's  $H(\text{curl}, \Omega)$ -conforming finite elements to discretize the problem. We present first results regarding the smoothing quality of the developed smoother for AMG.

**Key words.** AMG, Maxwell's equations, Nédélec elements, overlapping Schwarz smoother, distributive relaxation

16     **1. Introduction.** Algebraic Multigrid (AMG) methods [3] are a central tool in  
 17 the numerical solution of linear and nonlinear systems that arise from the discretiza-  
 18 tion of partial differential equations (PDEs). In this paper, AMG is applied to solve  
 19 the **2D definite Maxwell equation**,

$$\nabla \times \nabla \times u + \beta u = f, \quad \text{in } \Omega,$$

22 where  $\beta > 0$  is the spatially varying electrical conductivity,  $u$  is the unknown electric  
 23 field to be computed and  $f$  is the known right-hand side. The domain,  $\Omega$ , is an open,  
 24 bounded and connected Lipschitz domain in  $\mathbb{R}^2$ , and we impose Dirichlet boundary  
 25 conditions. Note that Equation (1) involves two 2D-curl operators, the 2D-curl of a  
 26 vector-valued function,  $w = (w_1, w_2)^T : \Omega \rightarrow \mathbb{R}^2$ , and the 2D-curl of a scalar function  
 27  $v : \Omega \rightarrow \mathbb{R}$ , defined as

$$\operatorname{curl} v := \begin{pmatrix} \partial_2 v \\ -\partial_1 v \end{pmatrix}, \quad \operatorname{curl} w := \partial_1 w_2 - \partial_2 w_1,$$

29 respectively.

30 The curl operators give rise to the standard Sobolev space,

$$H(\operatorname{curl}, \Omega) := \{v \in L^2(\Omega, \mathbb{R}^2) \mid \operatorname{curl} v \in L^2(\Omega)\},$$

33 where  $L^2$  denotes the space of square Lebesgue integrable functions. We therefore  
 34 discretize our model problem using Nédélec's  $H(\text{curl}, \Omega)$ -conforming finite elements  
 35 [7]. The resulting linear system is denoted by

$$36 \quad (2) \qquad \qquad \qquad Au = f,$$

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<sup>†</sup>University of Wuppertal, ([lisa.claus@math.uni-wuppertal.de](mailto:lisa.claus@math.uni-wuppertal.de), <https://www.hpc.uni-wuppertal.de/hochleistungsrechnen-swt/mitarbeiter/lisa-claus.html>).

38 with  $A = N + P$ , and where  $N$  is the discrete approximation of the weak form of the  
 39 curl-curl term in (1), and where  $P$  is the discrete approximation to the weak form of  
 40 the  $\beta$  term in (1). The difficulty of using AMG to solve Maxwell's equations results  
 41 from the kernel of the curl operator. Using the identity

42 (3) 
$$\nabla \times (\nabla \Phi) = 0,$$

44 we see that gradients of scalar functions,  $\Phi$ , lie within the kernel of the curl operator.  
 45 Thus, the kernel includes the gradients of all differentiable scalar functions  $\Phi$ . Since  
 46 the gradient of a smooth function is smooth and the gradient of an oscillatory func-  
 47 tion is oscillatory, it is obvious that the kernel contains both smooth and oscillatory  
 48 functions. For further details, consult [6]. The discrete null-space analogue of (3) is

49 (4) 
$$FG = \Theta,$$

51 where the matrix  $G$  is a discrete gradient operator and  $\Theta$  denotes the zero matrix. The  
 52 matrix  $G$  is trivial to construct. In particular, each row contains at most two nonzeros  
 53 (with value  $\pm 1$ ) and corresponds to an edge between two nodes of the associated nodal  
 54 mesh.

55 The classical AMG method fixes the relaxation method, usually, a pointwise  
 56 smoother, and enforces an efficient interplay with the coarse-grid correction. This  
 57 procedure is not sufficient for Maxwell's equations, since Maxwell's equations have a  
 58 large near null space and, therefore, contain high frequency components in the near  
 59 null space (see Fig. 1 and [6] for details). In order to smooth these oscillatory com-  
 60 ponents and approximate them appropriately on coarser grids, a different relaxation  
 61 scheme than a simple pointwise smoother is necessary. Efficient relaxation methods  
 62 for geometric multigrid methods for Maxwell's equations are proposed in [2, 5]. In  
 63 this paper, we construct AMG smoothing methods based on these relaxation schemes.

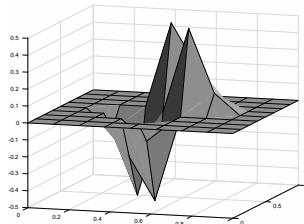


Fig. 1: High frequency component in the near null space for the 2D definite Maxwell equation on a quadrilateral grid.

65 This paper is structured as follows: Section 2 provides the definition of Nédélec  
 66 element discretization and we introduce the general idea of smoothing procedures for  
 67 Maxwell's equations. In Section 3, we describe two AMG smoothing methods for the  
 68 2D definite Maxwell equation. Numerical results are presented in Section 4. The  
 69 paper closes with a conclusion.

70 **2. Discretization and geometric multigrid for Maxwell's equations.** We  
 71 discretize the definite Maxwell equation (1) by linear **Nédélec elements**. Assuming  
 72 that  $\Omega$  is discretized by a quadrilateral mesh,  $\mathcal{T}$ , in 2D, Nédélec elements represent  
 73 basis functions in  $H(\text{curl}, \mathcal{T})$  spaces. We demonstrate the general approach to obtain  
 74 the system matrix  $A$  on the reference element, depicted in Figure 2.  
 75

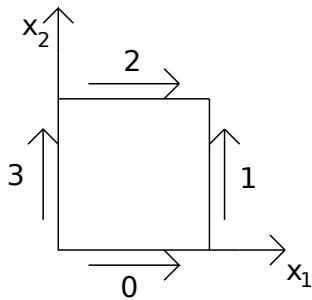


Fig. 2: The four degrees of freedom of 2D Nédélec elements in a reference configuration on a quadrilateral grid.

76 Let  $\eta$  denote the global edge basis functions and let  $x = (x_1, x_2)^T \in \Omega$  be a point.  
 77 The degrees of freedom (dofs),  $u$ , are related to the edges of the elements,

$$78 \quad 79 \quad \left\{ \alpha_i(u) = \int_{e_i} t_i \cdot u \, ds, \quad i \in \{1, 2, \dots\} \right\}$$

80 for every edge,  $e_i$ , in the reference element. There is one dof related to each edge,  
 81 therefore there are four dofs related to each element. The vector  $t_i$  is the tangential  
 82 unit vector of the edge  $e_i$ . One has to choose which direction for the unit tangential  
 83 vectors to use. Our choice is illustrated in Figure 2. The reference basis functions of  
 84 the Nédélec elements are given by the requirement  $\alpha_i(\eta_j) = \delta_{ij}$  [1, 7, 9],

$$85 \quad \eta_0(x) = \begin{pmatrix} 1 - x_2 \\ 0 \end{pmatrix}, \quad \eta_1(x) = \begin{pmatrix} 0 \\ x_1 \end{pmatrix}, \quad \eta_2(x) = \begin{pmatrix} x_2 \\ 0 \end{pmatrix}, \quad \eta_3(x) = \begin{pmatrix} 0 \\ 1 - x_1 \end{pmatrix}.$$

86  
 87 The global Nédélec basis functions are nonzero only in the two elements who  
 88 share the edge that is related to the basis function. A more detailed treatment can  
 89 be found in [7, 8].

90 **2.1. Geometric Multigrid smoothers.** Geometric multigrid with the overlapping  
 91 Schwarz smoother by Arnold, Falk and Winther (AFW) [2] or with distributive  
 92 relaxation proposed by Hiptmair [4, 5], is an efficient solver for Maxwell's equations.  
 93 We therefore construct two AMG smoothers based on these relaxation methods. The  
 94 goal of both smoothers is to treat the kernel with a special procedure to eliminate  
 95 the high frequency components. First, we review the idea of the geometric relaxation  
 96 procedures.

97  
 98 The geometric overlapping Schwarz smoother for Maxwell's equation is a block  
 99 smoother. The edge points are clustered into small overlapping blocks,  $\Omega_i$ , (see Figure 3)  
 100 and we solve the systems  $A_i u_i = f_i$ , where  $i$  indicates the number of the

102 current block,  $f_i$  is some right-hand side,  $u_i$  is a subvector which contains all un-  
 103 known edges/dofs of block  $\Omega_i$ , and  $A_i$  is a submatrix of the Maxwell matrix  $A$  that  
 104 belongs to block  $\Omega_i$ . The number of blocks corresponds to the number of nodal points  
 105 in the grid. The blocks contain all geometric nearest neighbor edges of the nodal  
 106 points, depicted in Figure 3.

107

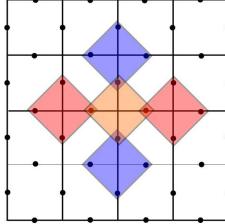


Fig. 3: Geometric idea of grid separation on quadrilateral grids.

108 Geometric distributive relaxation for Maxwell's equations uses a gradient matrix,  
 109  $G$ , that is constructed such that each row corresponds to an edge between two nodes  
 110 in the associated nodal mesh. Equivalently, the matrix  $G$  can be constructed such  
 111 that it uses the same blocks,  $\Omega_i$ , as the overlapping Schwarz smoother. As described  
 112 above, each block  $\Omega_i$ , defines a submatrix,  $A_i$ , of the Maxwell matrix,  $A$ . The eigen-  
 113 vectors corresponding to the smallest eigenvalue of each submatrix,  $A_i$ , define the  
 114 columns of the gradient matrix,  $G$ . Therefore, the number of columns of the matrix,  
 115  $G$ , equals the number of blocks,  $\Omega_i$ . The gradient matrix  $G$  is applied as a projector  
 116 for the Maxwell matrix,  $A$ , defining a new matrix,  $M = G^T A G$ , and a Gauß-Seidel  
 117 smoother (GS) is applied to this new matrix  $M$ .

118

119 **3. AMG smoothers.** In contrast to the geometric multigrid method, the geo-  
 120 metric structure of the dofs is unknown in the AMG setting. Instead, the dependen-  
 121 cies of the dofs are given by the matrix. Therefore, geometric smoothers need to be  
 122 adapted for AMG. Based on the geometric relaxation methods by AFW [2] and Hipt-  
 123 mair [4, 5], described in Section 2.1, we construct an overlapping Schwarz smoother  
 124 and a distributive relaxation for AMG.

125 **3.1. Overlapping Schwarz smoother for AMG.** The idea of the overlapping  
 126 Schwarz smoother is to separate the degrees of freedom into overlapping subsets,  $\Omega_i$ ,  
 127 and solve the corresponding systems,  $A_i u_i = f_i$ , as described in Section 2.1. Note  
 128 that the AMG setting influences the choice of the overlapping subsets,  $\Omega_i$ . We will  
 129 see that Maxwell's equations require the solution of all systems that contain a near  
 130 null space component and that correspond to subsets with a diameter smaller than  
 131 (or equal to) the coarsening factor,  $m$ . In this paper, the coarsening factor is chosen  
 132 to be two. The coarse grid handles the larger subdomains and near null space vectors.  
 133 After smoothing on all of these systems, GS is applied to the system defined by the  
 134 remaining points,  $\Omega_{\text{point}}$ .

135 A simple strategy for choosing the overlapping subsets  $\Omega_i$ , can be described as  
 136 follows:

**Algorithm 1** Select  $\Omega_i$ 


---

```

Set  $S := \emptyset$ ,  $\Omega_i := \{i\}$ ,  $\Omega_{\text{point}} = \Omega$ 
for all  $\Omega_i$  do
     $\Omega_i := \Omega_i \cup \{\text{distance 1 neighbors of } i\}$ 
     $A_i = A(\Omega_i, \Omega_i)$ 
    if  $A_i$  is nearly singular then
         $S \leftarrow S \cup \{\Omega_i\}$ 
         $\Omega_{\text{point}} \leftarrow \Omega_{\text{point}} - \Omega_i$ 
    end if
end for

```

---

137 For each  $i$ , starting with a predefined set,  $\Omega_i$  (in the above pseudocode,  $\Omega_i$  is  
 138 chosen such that it contains one edge dof), certain edge dofs are added to the sets,  
 139  $\Omega_i$ , (for example the distance 1 neighbors of the initial edges). Submatrices are  
 140 constructed from the system matrix,  $A$ , by restricting it to the columns and rows that  
 141 belong to the edge dofs in set  $\Omega_i$ . The last step during the loop is to check if the  
 142 submatrix includes a near null space component and at the end keep all sets for which  
 143 the submatrix includes a near null space component.

144 Remark: The above algorithm only illustrates the general idea for choosing sub-  
 145 sets,  $\Omega_i$ . In particular, considering only distance 1 neighbors is not sufficient to obtain  
 146 efficient AMG smoothers. An explanation for this is given in Section 3.3 along with  
 147 a different strategy that does produce good results.

148 After selecting the sets,  $\Omega_i$ , we apply the overlapping Schwarz algorithm:

**Algorithm 2** Overlapping Schwarz algorithm

---

```

for  $k = 1, 2, \dots$  do
    for  $\Omega_i \in S$  do
         $r_k \leftarrow f - Au_k$ 
         $u_k \leftarrow u_k + I_{\Omega_i} A_i^{-1} I_{\Omega_i}^T r_k$ 
    end for
     $r_k \leftarrow f - Au_k$ 
     $u_{k+1} = u_k + I_{\Omega_{\text{point}}} A_{\text{point}}^{\sim 1} I_{\Omega_{\text{point}}}^T r_k$ 
end for

```

---

149 Here,  $A, A_i, f$  and  $u$  are defined as introduced in Section 1 and Section 2.1. The  
 150 injection matrix with regard to a subdomain,  $d$ , is indicated by  $I_d$  and  $d = \Omega_{\text{point}}$  refers  
 151 to the set of remaining points, which are not included in one of the near null space sets  
 152  $\Omega_i$ . The term  $A_{\text{point}}^{\sim 1}$  denotes the approximate solution of the system  $A_{\text{point}} u_{\text{point}} =$   
 153  $I_{\Omega_{\text{point}}}^T r_k$  by the application of the Gauß-Seidel relaxation method. We use a forward  
 154 iteration loop for the presmoother and a backward iteration loop for the postsmoother  
 155 to obtain a symmetric algorithm.

156 **3.2. Distributive relaxation smoother for AMG.** The distributive relax-  
 157 ation smoother uses the subsets,  $\Omega_i$ , to construct a matrix,  $G$ , that is equivalent to  
 158 the discrete gradient matrix used in the geometric Hiptmair smoother described in  
 159 Section 2.1. The Gauß-Seidel relaxation method is used to relax all dofs and after-  
 160 wards the Galerkin matrix,  $G^T AG$ , is used to relax the near null space components  
 161 with fine scale support.

**Algorithm 3** Distributive Relaxation algorithm

---

```

for  $\Omega_i \in S$  do
  Compute smallest eigenvector  $v_i$  of  $A_i$ 
  Add  $v_i$  to columns of  $G$ 
end for
for  $k = 1, 2, \dots$  do
   $r_k \leftarrow f - Au_k$ 
   $u_k \leftarrow u_k + A^{-1}r_k$ 
   $r_k \leftarrow f - Au_k$ 
   $u_{k+1} = u_k + G(G^T AG)^{-1}G^T r_k$ 
end for

```

---

162 Here,  $A, A_i, G, f$  and  $u$  are defined as introduced in Section 1 and Section 2.1. The  
 163 terms  $(G^T AG)^{-1}$  and  $A^{-1}$  denote the approximate solution by the application of the  
 164 symmetric Gauß-Seidel relaxation method.

165 **3.3. Subdomain decision.** The general idea of the AMG smoothers is to eliminate  
 166 the near null space components that cannot be represented on the coarse grid.  
 167 This can be done in the following way:

168  
 169  $S = \emptyset$   
 170  $K = \text{all sets with diameter } \leq m$   
 171 **for**  $J \in K$  **do**  
 172 **for all**  $J$  that include near null space components **do**  
 173  $S \leftarrow S \cup \{J\}$   
 174 **end for**  
 175 **end for**

176 Since in this paper, we use a coarsening factor of  $m = 2$ , we inspect all sets with  
 177 diameter  $m = 2$  and check for near null space components. Then, we use one of the  
 178 two relaxation methods, overlapping Schwarz relaxation or distributive relaxation.

179  
 180 Sufficient subsets are necessary to eliminate near null space components. As  
 181 already mentioned in Section 3.1, the distance 1 neighbor sets are not sufficient for  
 182 AMG. In the context of geometric multigrid, nearest neighbors are closest edge dofs to  
 183 a nodal point in a geometrical sense (see Fig. 3). In the AMG context, we do not know  
 184 the location of nodal points and edge dofs. Instead, distance 1 neighbors in terms of  
 185 AMG are the closest edge dof to another edge dof (see Fig. 4). Closest edges to edge  $i$   
 186 in the AMG context are all edge dofs that have a corresponding matrix entry in row  $i$ .  
 187



Fig. 4: Examples of distance 1 neighbors in AMG context on quadrilateral grids.

188 The distance 1 neighbor sets in AMG are not sufficient as block smoothing sets  
 189  $\Omega_i$ , since they do not account for all diameter 2 subdomains. We need to establish  
 190 sets that cover all diameter 2 subdomains which include a near null space component.  
 191 Accordingly, we construct different sets that inspect all diameter  $m = 2$  subdomains  
 192 and includes near null space components. The algorithm looks as follows and is  
 193 explained below.

---

**Algorithm 4** Select  $\Omega_i$ 


---

```

Set  $S := \emptyset$ ,  $\Omega_{\text{point}} = \Omega$ 
for  $i = 1 : \# \text{dof}$  do
   $T = \{\text{all distance 2 neighbors of } i\}$ 
  for  $j \in T$  do
     $\Omega_{i,j} = \{i, j\} \cup \{\text{common distance 1 neighbors of } i \text{ and } j\}$ 
     $A_{i,j} = A(\Omega_{i,j}, \Omega_{i,j})$ 
    if  $A_{i,j}$  is nearly singular and  $\Omega_{i,j} \notin S$  then
       $S \leftarrow S \cup \{\Omega_{i,j}\}$ 
       $\Omega_{\text{point}} \leftarrow \Omega_{\text{point}} - \Omega_{i,j}$ 
    end if
  end for
end for

```

---

194 In the special case of quadrilateral elements (see Fig. 5), the algorithm above  
 195 generates the same sets as AFW and Hiptmair in the geometric case. For all degrees  
 196 of freedom, first the algorithm constructs a set that includes all distance  $m = 2$   
 197 neighbors (illustrated by crosses in Figure 5). Thereafter, a set,  $\Omega_{i,j}$ , is constructed  
 198 that includes the initial dof,  $i$ , (red rectangle in Fig. 5), a distance 2 neighbor,  $j$  and  
 199 all common distance 1 neighbors of  $i$  and  $j$ . All matrices are checked if they include  
 200 a near null space and we keep just the sets that include a near null space. So far, we  
 201 did not find a way to compute these sets in a more efficient way.

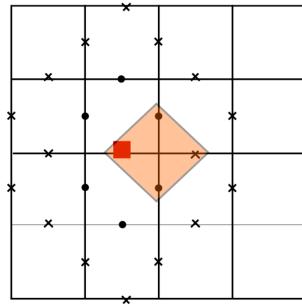


Fig. 5: Example of an AMG subdomain,  $\Omega_{ij}$ , constructed with Alg. 4. The red box indicates a given initial dof,  $i$ , crosses indicate distance 2 points of  $i$ , and dots indicate distance 1 points of  $i$ .

202 **4. Numerical Results.** The numerical results in this section show the efficiency  
 203 of the AMG overlapping Schwarz smoothing method (OSS) and the AMG distributive  
 204 relaxation method (DR) for the definite Maxwell equation.

205     **4.1. Cost comparison.** The algorithms in Section 3.1 and Section 3.2 both  
 206 have a setup phase to create the sets  $\Omega_i$ . In addition, DR generates the matrix  $G$  out  
 207 of the sets  $\Omega_i$ . OSS applies a block inverse for each near null space component, so it is  
 208 a bit more expensive in the solve phase. In contrast DR uses the Gauß-Seidel method  
 209 to solve the transformed system. OSS has the potential to be cheaper in the setup  
 210 phase because the actual near null space components do not need to be computed.  
 211 For example, a few iterations of a pointwise smoother can determine slow convergence  
 212 and hence the presence of a near null space.

213     **4.2. Performance.** First, we show numerical results of the smoother by itself  
 214 without an underlying multigrid method. For this purpose, we apply a Fourier mode  
 215 smoothing study for OSS and DR as described in Section 3.

216     The Fourier mode study can be described as follows: The homogenous system  
 217  $Au = 0$  is iteratively solved with different initial guesses. The system matrix  $A$  is  
 218 generated by using  $\beta = 0.1$ . The initial guesses,  $\sin(k\pi x_1) \sin(l\pi x_2)$ , with  $l = k =$   
 219  $1, \dots, n-2$  are the Fourier modes. Here  $x_1, x_2$  denote the coordinates of the unknowns  
 220 in the grid and  $n$  indicates the number of nodal gridpoints in one direction. We set  
 221  $l = k$  to obtain 2D results.

222     The results of the Fourier mode study for both smoothers are presented in Figure 6.  
 223 We indicate results for the 33x33 grid with 2112 dofs. We plot the number of  
 224 iterations needed to reduce the infinity norm of the error by 0.001 for every mode  
 225  $k$ . The number of iterations in Figure 6 is between 7 and 338 for DR and between 5  
 226 and 663 for OSS.

227     The initial guesses  $\sin(k\pi x_1) \sin(l\pi x_2)$  with small  $k$  indicate smooth functions,  
 228 whereas these initial guesses with large  $k$  indicate oscillatory functions. Good  
 229 smoothers should reduce oscillatory components efficiently.

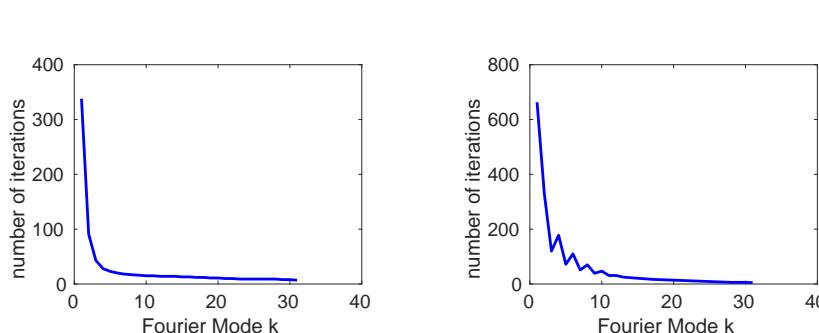


Fig. 6: Number of iterations for different Fourier modes with DR (left) and OSS (right).

231     In the following, we show numerical results of an AMG two-grid method using the  
 232 AMG overlapping Schwarz smoothing method and the AMG distributive relaxation  
 233 method as described in Section 3. The numerical results in this section demonstrate  
 234 the efficiency of the AMG smoothers for the definite Maxwell equation and we compare  
 235 both smoothers.

236     We consider the following model problem: take  $\Omega = [0, 1]^2$ ,  $\beta = 0.1$  and the right-  
 237 hand-side is set to zero. We use a random initial guess  $u_0(x, y)$ . The model problem is  
 238 solved with one pre- and one post smoothing step ( $V(1, 1)$ ). Since it is not the concern  
 239 of this paper to find an appropriate AMG coarsening, we construct the coarse grid

240 according to the geometric multigrid, see Figure 7. The coarse grid points are new  
 241 points on the grid and not a subset of the fine grid points. We use a restriction with  
 242 the weighting,

$$243 \quad \begin{bmatrix} \frac{1}{4} & \frac{1}{4} \\ \frac{1}{2} & \bullet \\ \frac{1}{4} & \frac{1}{4} \end{bmatrix}, \quad \begin{bmatrix} \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \\ \bullet & & \\ \frac{1}{4} & \frac{1}{2} & \frac{1}{4} \end{bmatrix},$$

244 where  $\bullet$  denotes the resulting coarse grid point. The left stencil denotes the restriction  
 245 of an x-edge coarse point and the right stencil of a y-edge coarse point. The interpo-  
 246 lation matrix is the transpose of the restriction matrix and we use the Galerkin coarse  
 247 grid operator.

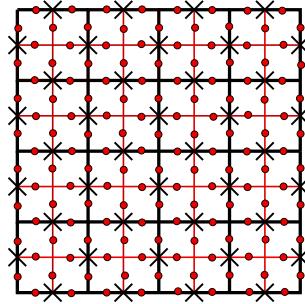


Fig. 7: The coarse grid (black) and the corresponding coarse grid points  $\times$  resulting from the fine grid (red) and the fine grid points  $\bullet$ .

248 The convergence rate of the AMG two-grid method for different grid sizes and  
 249 both smoothers is shown in the following. We show the results of three different grid  
 250 sizes, 17x17 grid with 544 degrees of freedom, 33x33 grid with 2112 dofs and 65x65  
 251 grid with 8320 dofs. We use the 2-norm to measure the convergence.

252 The asymptotic convergence factors of the two-level AMG method with OSS and  
 253 DR are indicated in Table 1. These results present an asymptotic convergence factor  
 254 of 0.19 for the two-level method with OSS and an asymptotic convergence factor of  
 255 0.14 for the two-level method with DR. We observe h-independent convergence rates.  
 256 Comparable results can also be found in [6].

#dofs	OSS	DR
544	0.19	0.13
2112	0.19	0.13
8320	0.19	0.14

Table 1: Convergence factors,  $\|r_k\|_2/\|r_{k-1}\|_2$ ,  $k \rightarrow \infty$ , of the two-level AMG with OSS and DR.

257 The results of the AMG method with DR look more promising. However, the DR  
 258 uses an expensive symmetric GS method for the entire system  $Au = f$ , whereas the  
 259 OSS is symmetrized by forward pre-relaxation and backward post-relaxation. Using  
 260 this approach for DR yields an asymptotic convergence rate of 0.37.

5. **Conclusion.** We developed AMG based smoothers for the definite Maxwell equation, the overlapping Schwarz smoother and distributive relaxation. The results of the two-grid methods with OSS and DR for the definite Maxwell equation are comparable with results of already known methods for geometric multigrid which can be found in the literature, cf. [6]. Analyzing the smoothers on more general problems is future work.

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