

# A fast direct solver for structured matrices

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## Introduction

We have developed a fast direct solver for structured linear systems based on multilevel matrix compression. Starting with a hierarchically block-separable matrix [2], we embed an approximation of the original matrix into a larger, but highly structured sparse one. The resulting representation allows for efficient storage, fast matrix-vector multiplication, fast matrix factorization, and fast application of the inverse.

The algorithm proceeds in two phases: a precomputation phase, consisting of matrix compression and factorization, followed by a solution phase to apply the matrix inverse. For boundary integral equations which are not too oscillatory, e.g., based on the Green's functions for the Laplace or low-frequency Helmholtz equations, both phases typically have complexity  $\mathcal{O}(N)$  in two dimensions, where  $N$  is the number of discretization points. In our current three-dimensional implementation, the corresponding costs are  $\mathcal{O}(N^{3/2})$  and  $\mathcal{O}(N \log N)$  for precomputation and solution, respectively. Extensive numerical experiments show a speedup of  $\sim 100$  for the solution phase over modern fast multipole methods; however, the cost of precomputation remains high. Thus, the solver is particularly suited to problems where large numbers of iterations would be required.

- (1) Reduced sensitivity to the conditioning of  $A$ .
- (2) Fast application of  $A^{-1}$  to multiple right-hand sides.
- (3) Efficient handling of low-rank perturbations of  $A$ .

Several closely related efforts:

- $H$ -matrices (Hackbusch et al.)
- HSS matrices (Gu, Chandrasekaran, et al.)
- skeletonization-based schemes (Martinsson and Rokhlin, G-, Gueyffier, Martinsson, Rokhlin).

## Structured matrices

Let  $A\mathbf{x} = \mathbf{f}$  be written in the form

$$\begin{aligned} A_{11}\mathbf{x}_1 + A_{12}\mathbf{x}_2 + \cdots + A_{1p}\mathbf{x}_p &= \mathbf{f}_1 \\ A_{21}\mathbf{x}_1 + A_{22}\mathbf{x}_2 + \cdots + A_{2p}\mathbf{x}_p &= \mathbf{f}_2 \\ &\vdots \\ A_{p1}\mathbf{x}_1 + A_{p2}\mathbf{x}_2 + \cdots + A_{pp}\mathbf{x}_p &= \mathbf{f}_p, \end{aligned} \quad (1)$$

where  $\mathbf{x}_i$  and  $\mathbf{f}_i$  are complex vectors of dimension  $N_i$  and  $A_{ij} \in \mathbb{C}^{N_i \times N_j}$ . Assuming  $N = \sum_{i=1}^p N_i$ , solution of the full linear system by classical Gaussian elimination is well-known to require  $\mathcal{O}(N^3)$  work.

**Definition** The matrix  $A$  is said to be *block-separable* [2] if each off-diagonal submatrix  $A_{ij}$  can be decomposed as the product of three low-rank matrices:

$$A_{ij} = L_i S_{ij} R_j \quad (2)$$

where  $L_i \in \mathbb{C}^{N_i \times k_i^r}$ ,  $S_{ij} \in \mathbb{C}^{k_i^r \times k_j^c}$ , and  $R_j \in \mathbb{C}^{k_j^c \times N_j}$ , with  $k_i^r \ll N_i$  and  $k_j^c \ll N_j$ . Note that in (2), the left matrix  $L_i$  depends only the index  $i$  and the right matrix  $R_j$  depends only on the index  $j$ .

$$A = D + LSR, \quad (3)$$

where

$$D = \begin{bmatrix} A_{11} & & \\ & \ddots & \\ & & A_{pp} \end{bmatrix} \quad (4)$$

is a block-diagonal  $N \times N$  matrix consisting of the diagonal blocks of  $A$ , and

$$S = \begin{bmatrix} 0 & S_{12} & \cdots & S_{1p} \\ S_{21} & 0 & \cdots & S_{2p} \\ \vdots & \vdots & \ddots & \vdots \\ S_{p1} & S_{p2} & \cdots & 0 \end{bmatrix} \quad (5)$$

is a dense  $K_r \times K_c$  matrix, where  $K_r = \sum_{i=1}^p k_i^r$  and  $K_c = \sum_{i=1}^p k_i^c$ , with zero diagonal blocks.  $L$  and  $R$  are given by

$$L = \begin{bmatrix} L_1 & & \\ & \ddots & \\ & & L_p \end{bmatrix}, \quad R = \begin{bmatrix} R_1 & & \\ & \ddots & \\ & & R_p \end{bmatrix}. \quad (6)$$

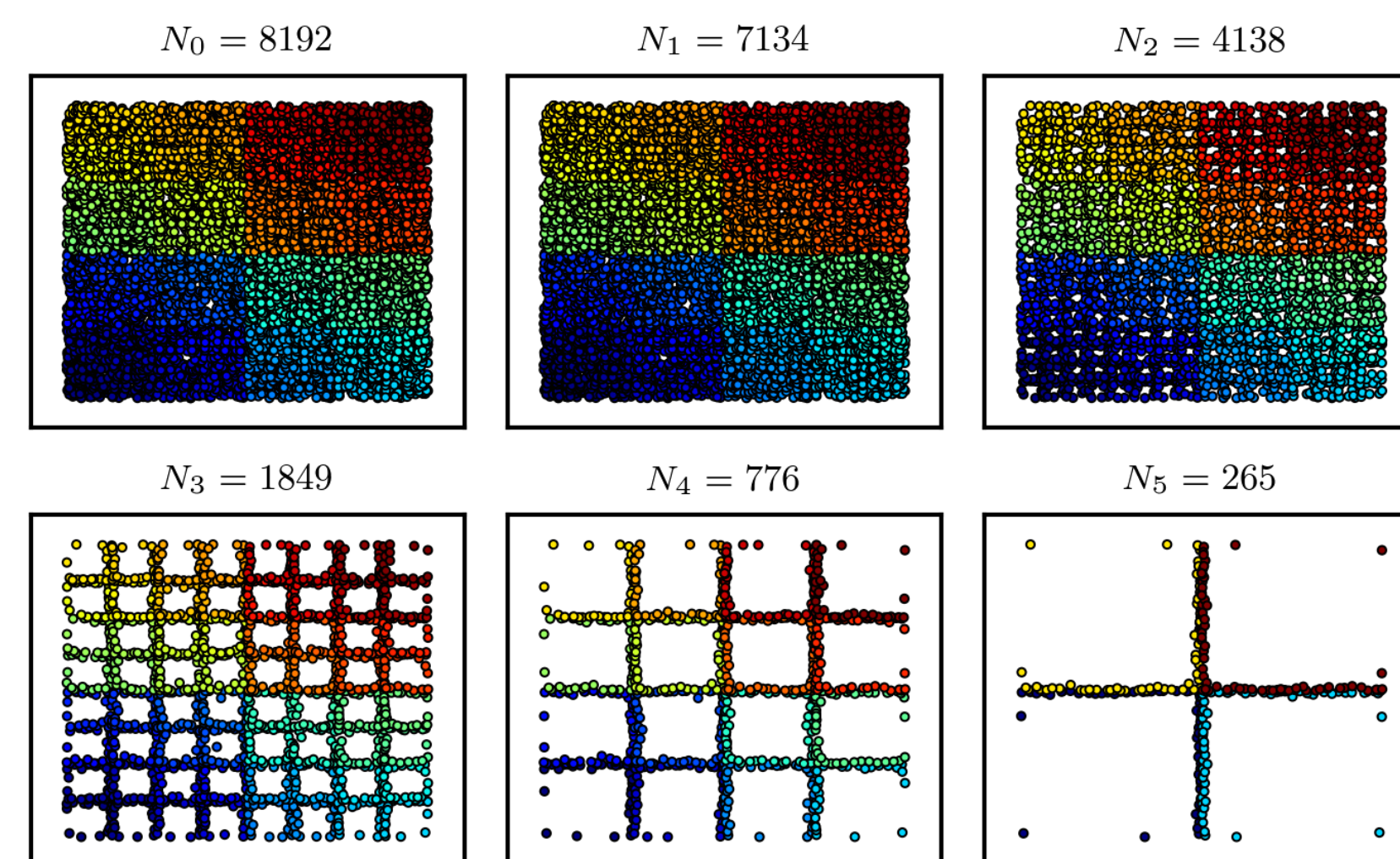
They are block-diagonal  $N \times K_r$  and  $K_c \times N$  matrices, respectively. This is, of course, an effective compression of  $A$  only when  $K_r, K_c \ll N$ . A useful feature of the representation (3) is that it permits rapid inversion. To see this, let  $z \equiv Rx$  and let  $y \equiv Sz$ . We can then write the original system in the form

$$\begin{bmatrix} D & L & \\ R & & -I \\ & -I & S \end{bmatrix} \begin{bmatrix} x \\ y \\ z \end{bmatrix} = \begin{bmatrix} b \\ 0 \\ 0 \end{bmatrix}. \quad (7)$$

Such a factorization can be computed recursively if the matrix supports the block-separable property at each level of the hierarchy. The result is a telescoping representation

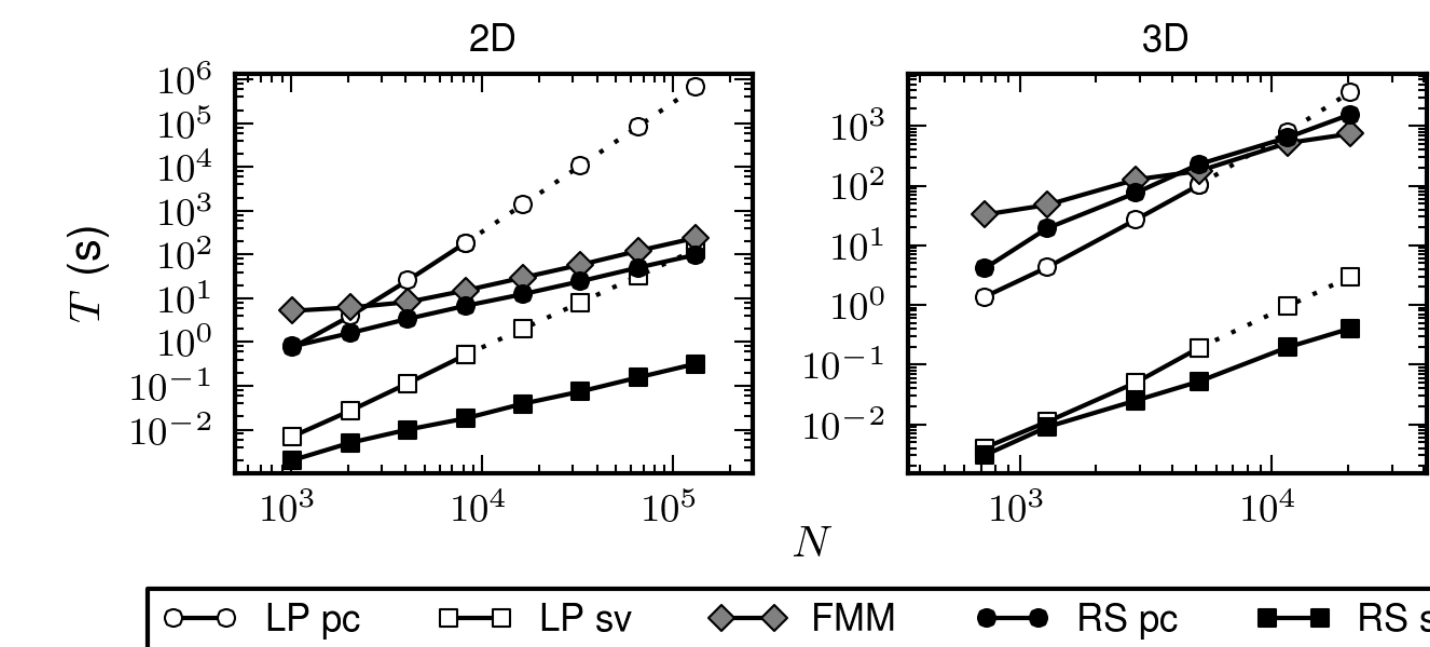
$$A \approx D^{(1)} + L^{(1)} \left[ D^{(2)} + L^{(2)} \left( \cdots D^{(\lambda)} + L^{(\lambda)} S R^{(\lambda)} \cdots \right) R^{(2)} \right] R^{(1)}$$

Example: Suppose  $\phi(x) = \int K(x, y) \rho(y) dy$  with  $K(x, y) = \log \|x - y\|$ , discretized at  $N = 8192$  points in the unit square, and compress to relative precision  $\epsilon = 10^{-3}$  using a five-level quadtree-based scheme.



Data sparsification by recursive skeletonization. At each level, the surviving skeletons are shown, colored by block index, with the total number of skeletons remaining given by  $N_i$ .

## Numerical results



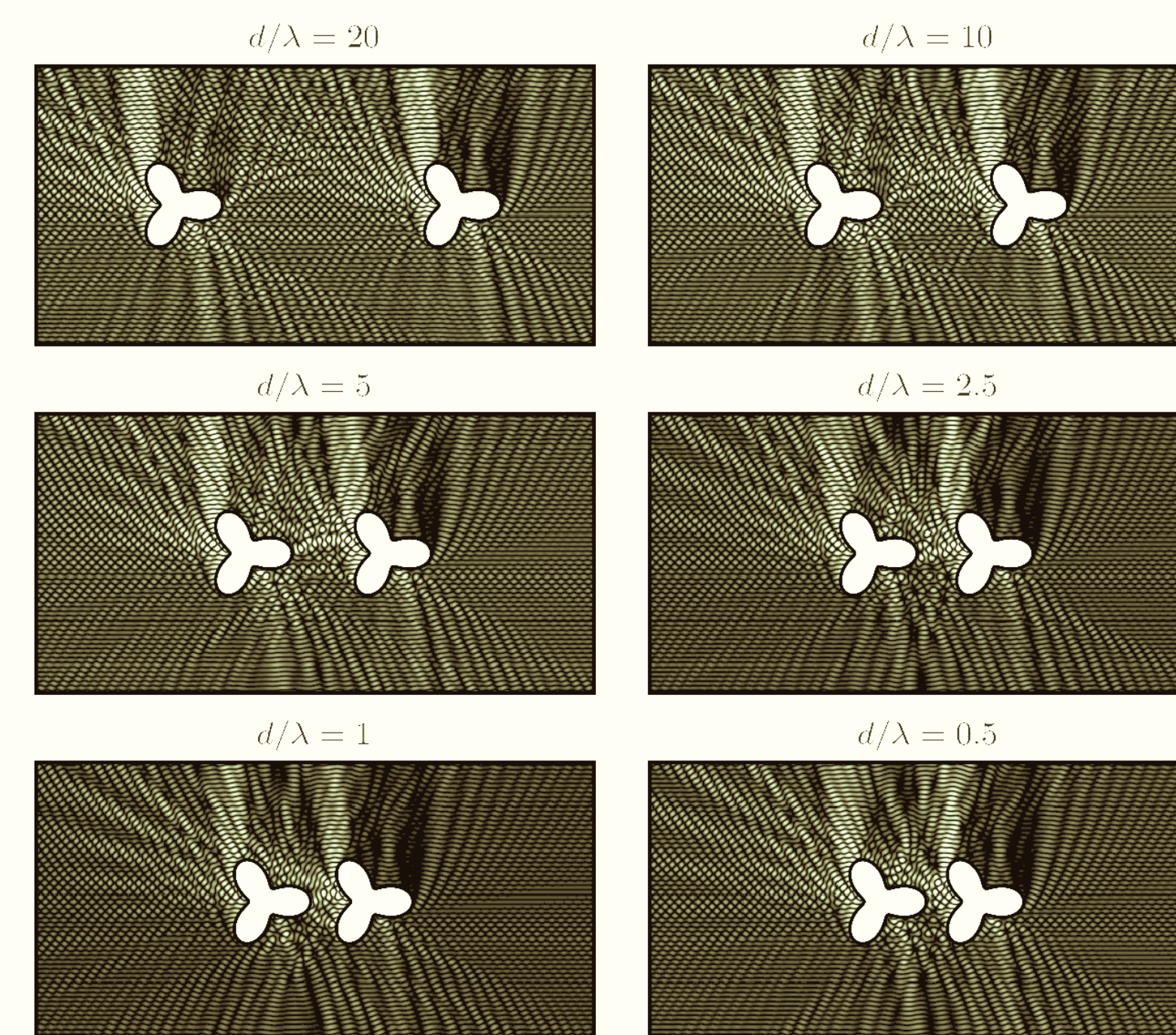
CPU times for solving the Helmholtz equation in various cases at low frequency ( $\omega = 10$  in 2D and  $\omega \approx 3.18$  in 3D) using LAPACK/ATLAS, FMM/GMRES, and recursive skeletonization; The precision was set to  $\epsilon = 10^{-9}$  in 2D and  $\epsilon = 10^{-6}$  in 3D.

## Multiple scattering

We show how our solver can be combined with fast iterative methods to great effect by considering a multiple scattering problem.

$$(\Delta + k^2)u = 0 \quad \text{in } \mathbb{R}^2 \setminus \bigcup_{i=1}^p \Omega_i$$

with, e.g., Neumann (sound-hard) boundary conditions



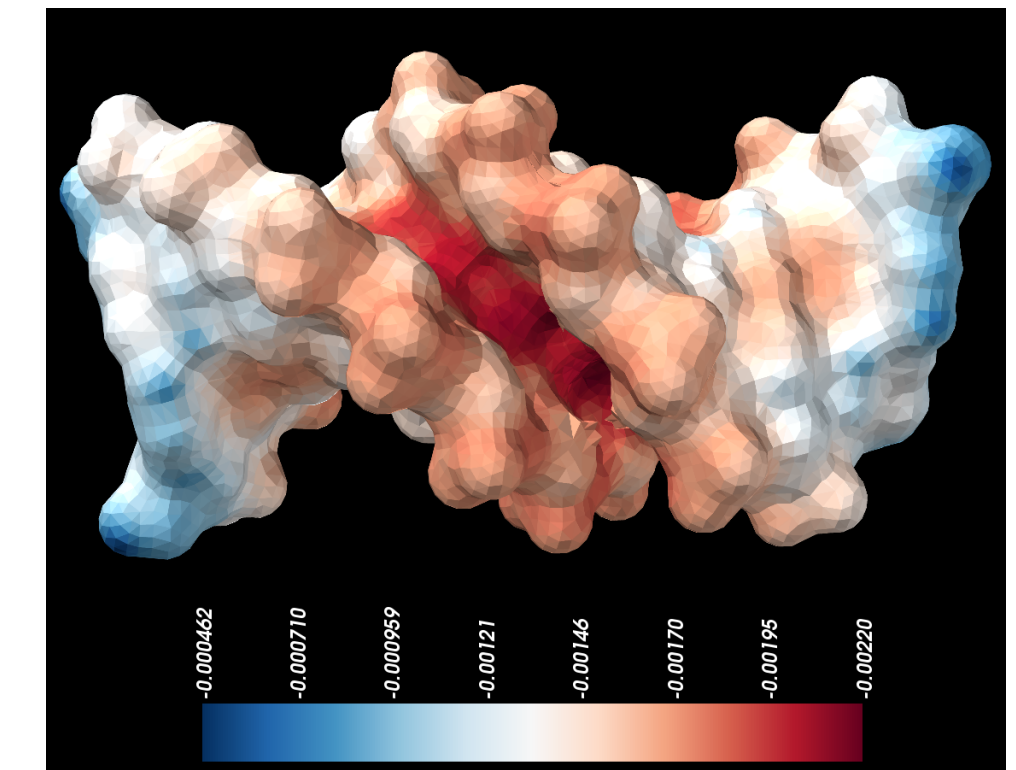
Real part  $\Re(u)$  of the pressure field in response to a vertical plane wave for various multiple scattering geometries characterized by the separation distance  $d/\lambda$  in wavelengths.

## Molecular electrostatics

Letting  $\Sigma$  be a molecular surface, a classical model for the electrostatic potential  $\varphi$  is Poisson's equation:

$$-\nabla \cdot [\varepsilon(x) \nabla \varphi(x)] = \sum_{i=1}^n q_i \delta(x - x_i), \quad (8)$$

We generated a molecular surface for a short segment of DNA consisting of  $N = 19752$  triangles and solved an integral equation for the induced polarization charge to precision  $\epsilon = 10^{-3}$ .



The total solution time was  $\approx 592$  s, with an inverse application time of  $\approx 0.08$  s, to be compared with  $\approx 27$  s using FMM/GMRES.

## Conclusions

We have developed a multilevel matrix compression and solution algorithm. The matrix structure required is fairly general and relies only on the assumption that the matrix have low-rank off-diagonal blocks. It is competitive with fast iterative methods based on FMM/GMRES in both 2D and 3D, provided that the kernel of the integral equation is not too oscillatory.

A principal limitation of all current hierarchical direct solvers is the growth in skeleton size in 3D, which prohibits the scheme from achieving optimal  $\mathcal{O}(N)$  complexity. New methods to curtail this growth are under development, at least for non-oscillatory cases.

Finally, although all numerical results are reported here for a single CPU core, the algorithm is naturally suited for HPC systems.

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