



Paraxial Approximation in CSR Modeling Using the Discontinuous Galerkin Method

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Introduction and Outline

- Well-established paraxial approximation in CSR has been previously modeled with a finite difference (FD) method.
- Novel approach in applying a discontinuous Galerkin (DG) method to the same equations.
- Starting point: nonhomogeneous Schrödinger type equations which arise from a paraxial approximation to Maxwell's equations.
- Include brief overview of DG and its implementation into our algorithm.
- Apply DG on a 2D transverse (x, y) grid and evolve the equations along the arclength s .
- Estimate the errors of our results and compute the longitudinal impedance.
- Discuss CPU vs GPU performance.
- Outline possible future work and new applications for DG.

Mathematical Problem

Domain of problem.

Consider rectangular domain with $(x, y) \in \Omega = [-a, a] \times [-b, b]$ and $s \in [0, L]$. Additional parameters satisfy: $k \in \mathbb{R}$ and $\rho > 0$.

Paraxial PDEs for $E^r(x, y, s; k)$:

$$\partial_s E_x^r = \frac{i}{2k} \nabla_{\perp}^2 E_x^r + \frac{ikx}{\rho} E_x^r + \frac{ikx}{\rho} E_x^b(x, y) \quad (1a)$$

$$\partial_s E_y^r = \frac{i}{2k} \nabla_{\perp}^2 E_y^r + \frac{ikx}{\rho} E_y^r + \frac{ikx}{\rho} E_y^b(x, y) \quad (1b)$$

$$E_s^r = \frac{i}{k} (\partial_x E_x^r + \partial_y E_y^r) \quad (1c)$$

Known beam component of field E^b :

$$E_x^b = C \frac{x}{x^2 + y^2}, \quad E_y^b = C \frac{y}{x^2 + y^2} \quad (2)$$

Boundary conditions for $E_{x,y}^r(x, y, s; k)$:

$$\begin{aligned} \partial_x E_x^r &= \partial_y E_y^b, & \text{on } x = \pm a \\ E_x^r &= -E_x^b, & \text{on } y = \pm b, \end{aligned} \quad (3)$$

$$\begin{aligned} E_y^r &= -E_y^b, & \text{on } x = \pm a \\ \partial_y E_y^r &= \partial_x E_x^b, & \text{on } y = \pm b. \end{aligned} \quad (4)$$

Initial conditions for $E_{x,y}^r(x, y, s; k)$:

$$\nabla_{\perp}^2 E_x^r = 0, \quad \nabla_{\perp}^2 E_y^r = 0, \quad \text{at } s = 0 \quad (5)$$

Physical Problem

Beam Setup.

Consider a line charge moving at $v \approx c$, on a circular arc of radius ρ and length L , with perfectly conducting rectangular vacuum chamber.

Beam coordinate system.

Adopt special case with line charge reduced to a single point. Maxwell's equations written in beam coordinates (x, y, s) where the arc is in the (x, s) plane, s is the distance along the arc, and (x, y) are perpendicular to arc.

Electric field transformation.

Relate the electric field $\mathcal{E}(x, y, s, t)$ to frequency domain field $\mathbf{E}(x, y, s; k)$ by Fourier-type transformation:

$$\mathcal{E}(x, y, s, t) \propto \int_{-\infty}^{\infty} dk \mathbf{E}(x, y, s; k) e^{ik(s-t)}$$

Initial condition of the electric field.

Assume fields reached steady-state from infinite straight prior to entering bend. Decompose electric field \mathbf{E} into two components: radiation and beam field denoted by \mathbf{E}^r and \mathbf{E}^b respectively. \mathbf{E}^b reduces the effect of the singularity.

DG Overview

- DG shares similarities with the finite element and finite volume methods.
- Rescale to dimensionless variables:

$$x \rightarrow ax, \quad y \rightarrow ay, \quad s \rightarrow 2ka^2s, \quad E_{x,y}^r \rightarrow Cu/a$$

- Equations (1a) and (1b) become:

$$\begin{aligned} -i\partial_s u &= \partial_x q_x + \partial_y q_y + F \\ q_x &= \partial_x u, \quad q_y = \partial_y u \end{aligned} \quad (6)$$

- Split Ω into K elements, select single element D .
- Local solution $u \in P^N(D)$.
- Multiply (6) by test functions $v \in P^N(D)$, integrate by parts over D , adjust boundary terms for fluxes (u^*, q_x^*, q_y^*) , and integrate by parts again:

$$\begin{aligned} -i \int_D dA (v \partial_s u) &= \int_D dA v (\partial_x q_x + \partial_y q_y + F) \\ &- \int_{\partial D} dL v [n_x (q_x - q_x^*) + n_y (q_y - q_y^*)] \end{aligned} \quad (7a)$$

$$\begin{aligned} \int_D dA (v q_{x,y}) &= \int_D dA v (\partial_{x,y} u) \\ &- \int_{\partial D} dL v n_{x,y} (u - u^*) \end{aligned} \quad (7b)$$

- Expand local solution in nodal Lagrange basis:

$$u(x, y) = \sum_{j=1}^{N_p} u_j \ell_j(x, y), \quad v(x, y) = \ell_i(x, y)$$

- Vectors for nodal values $\mathbf{u} = (u_1, u_2, \dots, u_{N_p})^T$ with $N_p = (N+1)(N+2)/2$.
- Similarly for $q_{x,y}, F, \ell$, then (7a)-(7b) become:

$$\begin{aligned} -iM \frac{d\mathbf{u}}{ds} &= \mathcal{S}_x \mathbf{q}_x + \mathcal{S}_y \mathbf{q}_y + \mathcal{M} \mathbf{F} \\ &- \int_{\partial D} dL n_x (\mathbf{q}_x - \mathbf{q}_x^*) \ell \end{aligned} \quad (8a)$$

$$- \int_{\partial D} dL n_y (\mathbf{q}_y - \mathbf{q}_y^*) \ell$$

$$\mathcal{M} \mathbf{q}_x = \mathcal{S}_x \mathbf{u} - \int_{\partial D} dL n_x (\mathbf{u} - \mathbf{u}^*) \ell \quad (8b)$$

$$\mathcal{M} \mathbf{q}_y = \mathcal{S}_y \mathbf{u} - \int_{\partial D} dL n_y (\mathbf{u} - \mathbf{u}^*) \ell \quad (8c)$$

- Mass and stiffness matrices:

$$\mathcal{M}_{ij} = \int_D dA \ell_i \ell_j, \quad (\mathcal{S}_{x,y})_{ij} = \int_D dA (\partial_{x,y} \ell_i) \ell_j$$

- Numerical fluxes depend on adjacent elements:

$$q_{x,y}^* = \{ \{ q_{x,y} \} \} - \tau [u]_{x,y}, \quad u^* = \{ \{ u \} \}$$

- Full details found in *Nodal Discontinuous Galerkin Methods*, (New York: Springer, 2008) by J. Hesthaven and T. Warburton.

DG Algorithm

Step 1: Build Elements and Matrices

- Partition Ω into $K = 2N_x^{res} N_y^{res}$ triangles and space nodes optimally for matrix conditioning.

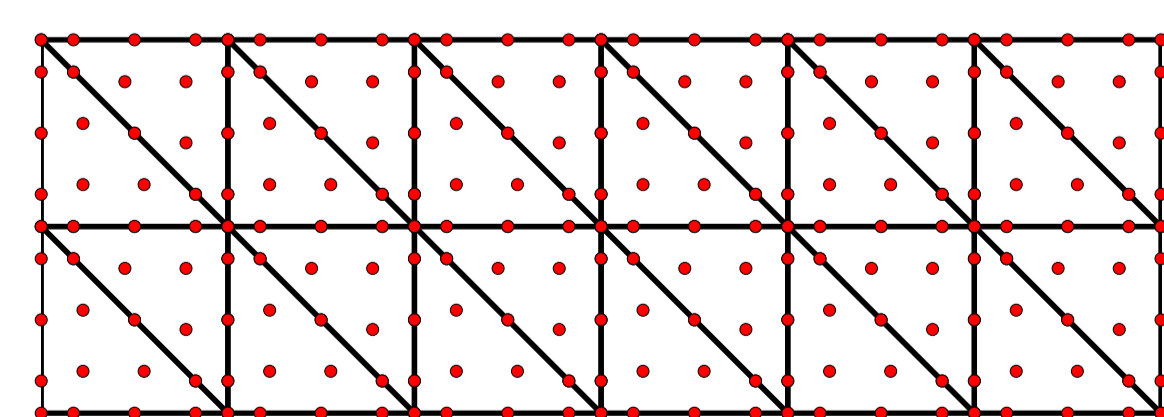


Figure 1: Mesh of $N_x^{res} = 6, N_y^{res} = 2, N = 4$.

- Build collocation derivative matrices: $\mathcal{D}_x = \mathcal{M}^{-1} \mathcal{S}_x$ and $\mathcal{D}_y = \mathcal{M}^{-1} \mathcal{S}_y$.

Step 2: Compute Initial Conditions

- Use sparse DG Poisson solver on (5) with (3), (4) to obtain initial $E_{x,y}^r$.
- Generate initial E_s^r with derivative matrices:

$$E_s^r = \frac{i}{k} (\mathcal{D}_x E_x^r + \mathcal{D}_y E_y^r) \quad (10)$$

DG Algorithm Cont.

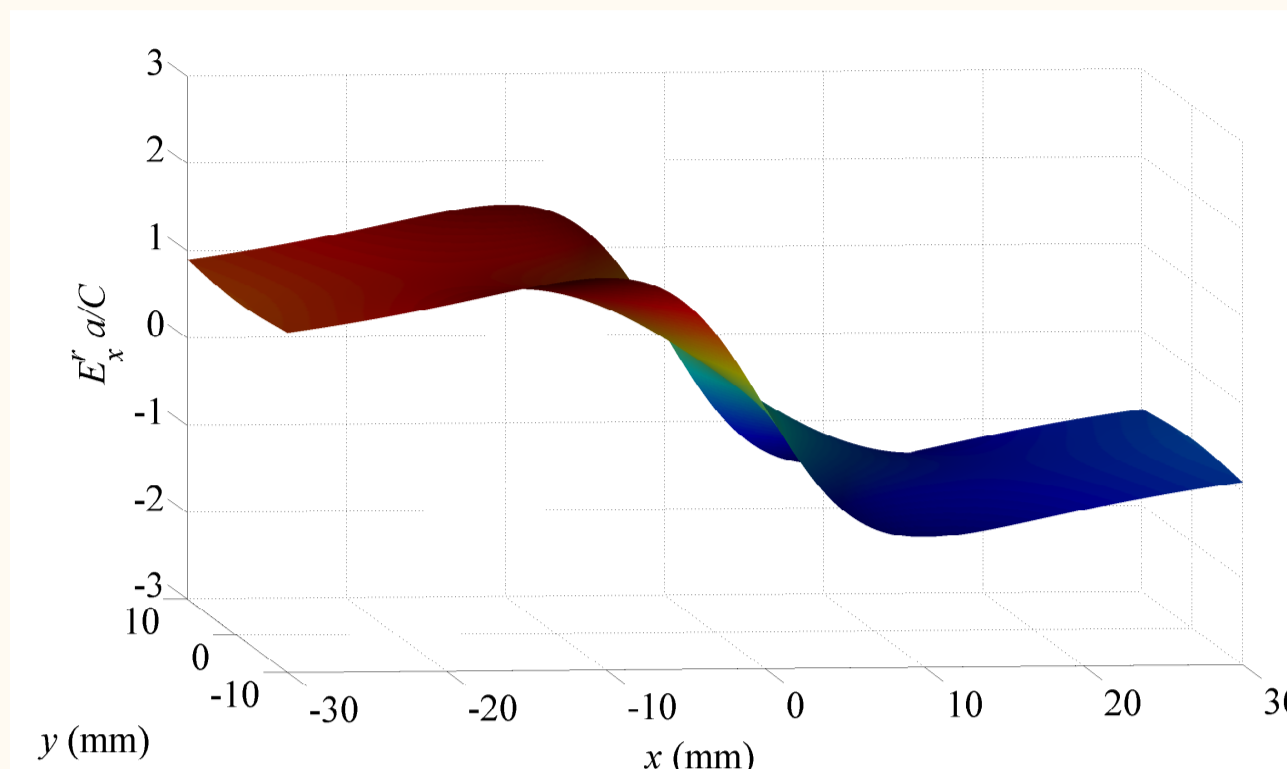


Figure 2: E_x^r initial state prior to entering bend.

Step 3: Evolve the Fields

- Estimate step size for evolution by:

$$\Delta s = C_s \cdot k \cdot r_{min}^2 \quad (11)$$

r_{min} is the minimum distance between all nodes, C_s is CFL-like constant of $\mathcal{O}(1)$. Note: $r_{min} \propto 1/(KN^2)$.

- Compute $\mathbf{q}_{x,y}$ with (8b)-(8c) and insert into (8a) for right-hand-side of $d\mathbf{u}/ds$.

- Use 4th order explicit Runge-Kutta to evolve.

- At each step, compute E_s^r with (10).

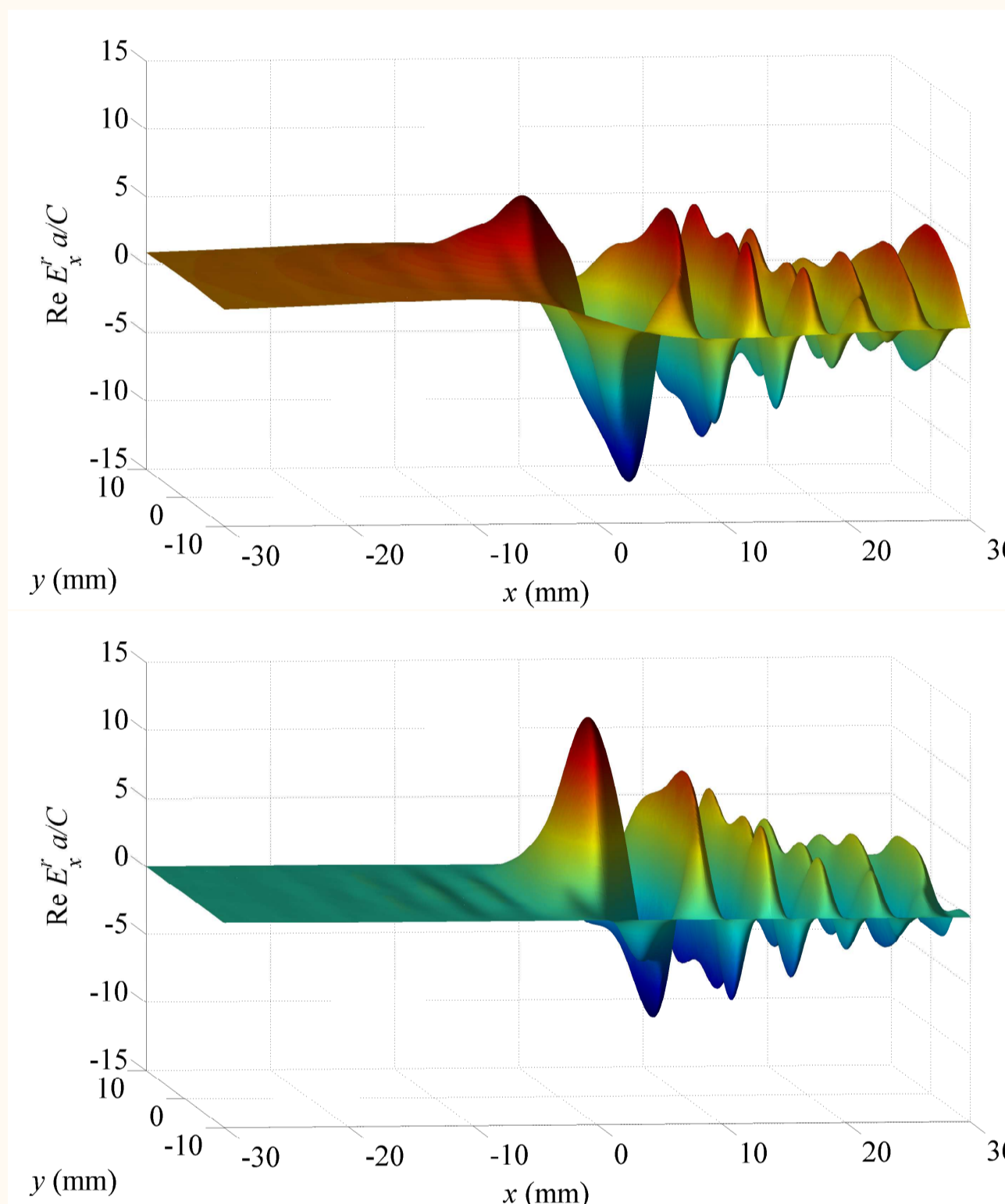


Figure 3: Real (top) and imaginary (bottom) parts of E_x^r after bend for $a = 30\text{mm}, b = 10\text{mm}, L = 200\text{mm}, \rho = 1\text{m}$, and $k = 8\text{mm}^{-1}$.

Step 4: Compute Impedance

- Evaluate impedance Z in two parts:

$$\begin{aligned} Z &= Z_b + Z_s \\ Z_b &= -\frac{Z_0}{2\pi C} \int_0^L ds E_s(0, 0, s; k) \\ Z_s &= -\frac{Z_0}{2\pi C} \int_L^\infty ds E_s(0, 0, s; k) \end{aligned} \quad (12)$$

- For details on the evaluation of Z , see D. Zhou's paper: Jpn. J. Appl. Phys. 51 016401 (2012).

Numerical Results: FD

- FD MATLAB code is CPU-only based on Agoh and Yokoya: PR-STAB 7 054403 (2004) which uses 2nd-order stencil with leap-frog evolution.

FD E_x^r Error and Computation Time				
Grid	61×21	121×41	181×61	241×81
	3.845e-1	1.126e-1	4.016e-2	3.440e-2
	4.539e-1	1.223e-1	4.551e-2	2.840e-2
	8s	125s	642s	2032s
	200	800	1800	3200

Table 2: Point-wise (top), L^2 error (upper middle), computation times (lower middle), and number of timesteps (bottom) for FD code using CPU.

Impedance Comparisons

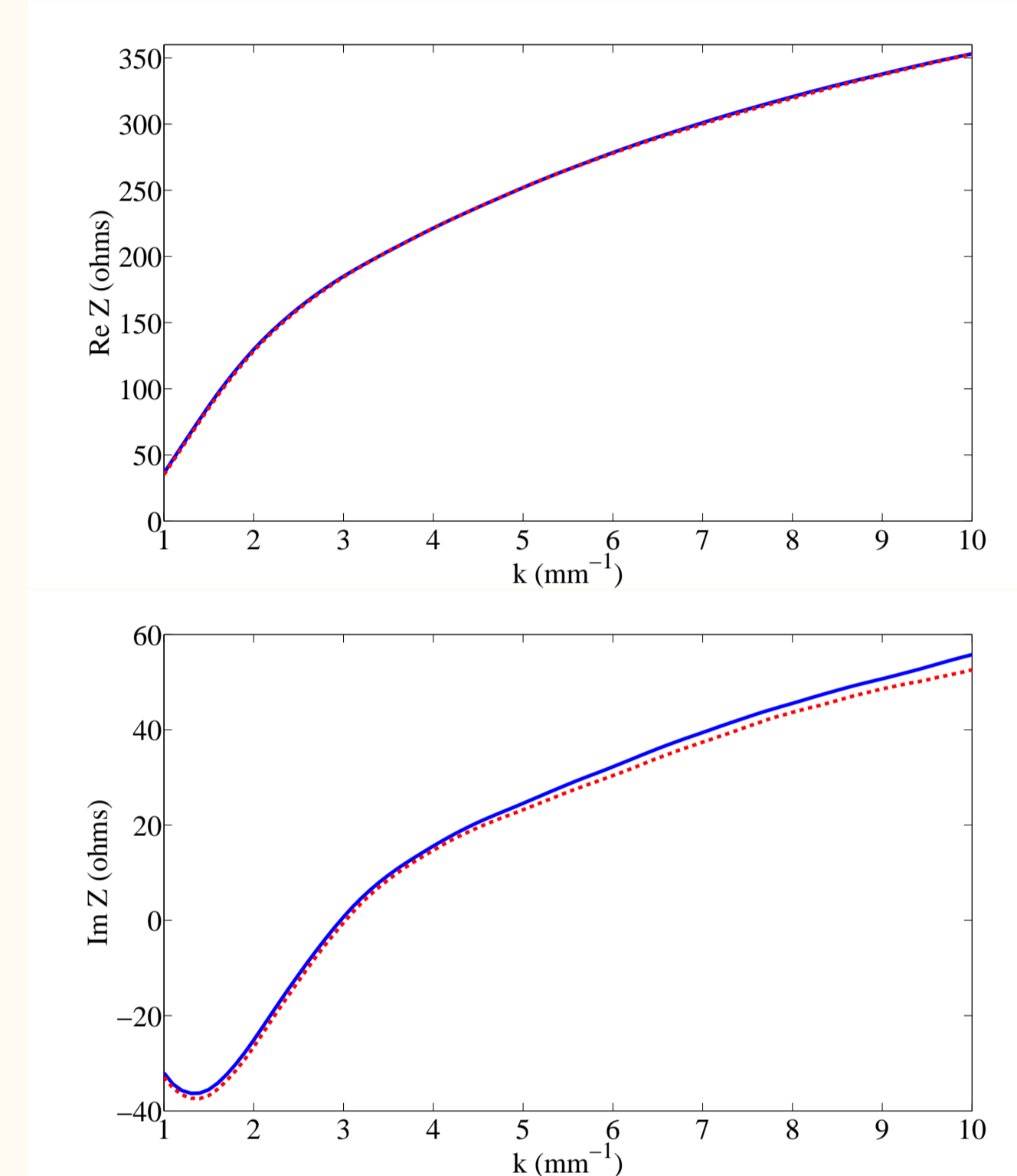


Figure 4: Real (top) and imaginary (bottom) parts of the longitudinal impedance [DG (blue solid), FD (red dashed)] for $a = 30\text{mm}, b = 10\text{mm}, L = 200\text{mm}, \rho = 1\text{m}$, and $k = 8\text{mm}^{-1}$. Agrees with D. Zhou referenced above.

GPU Computing Comments

- GPU code was adapted from CPU code with MATLAB's `gpuArray` CUDA kernel.
- MATLAB's GPU computing scales favorably for larger problems with large matrix-matrix operations or highly parallel tasks.
- Observed $\sim 60\%$ GPU usage for high resolution run with matrices of size 45×2400 .
- CPU : Intel Xeon E5-1620 ($\sim 80\text{Gflops/sec}$)
- GPU : NVIDIA GTX Titan ($\sim 1.6\text{Tflops/sec}$)

Numerical Results: DG

DG E_x^r Error and Computation Time				
$N \setminus K$	150	600	1350	2400
2	8.107e-1	3.915e-1	7.471e-2	2.453e-2
	8.032e-1	2.528e-1	4.291e-2	1.484e-2
	9s	28s	49s	81s
4	1.265e-1	4.897e-3	9.344e-4	3.590e-4
	8.017e-2	3.427e-3	9.462e-4	4.342e-4
	29s	88s	202s	392s
6	539	2156	4850	8622
	1.122e-2	5.177e-4	1.407e-4	5.483e-5
	6.974e-3	6.187e-4	1.932e-4	8.045e-5
8	83s	283s	677s	1319s
	1691	6764	15218	27054
	1.569e-3	1.672e-4	5.612e-5	$N \setminus A^*$
8	1.493e-3	2.098e-4	7.473e-5	$N \setminus A^*$
	208s	723s	1867s	3630s
	4174	16693	37559	66771

*.Used for comparison to other tests.

Table 1: Point-wise (top), L^2 error (upper middle), computation times (lower middle), and number of timesteps (bottom) for DG code using MATLAB `gpuArray` implementation.

Future Work

- Examine spectral convergence order for DG.
- Design a higher order FD code with GPU implementation and compare performance with DG.
- Explore possible perturbation expansion of $2k^2 a^3 / \rho$ in rescaled versions of (1a) and (1b).
- Implement DG on Maxwell's equations without paraxial approximation and compare results.
- Consider DG on Vlasov-Maxwell's equations for future applications.

Acknowledgments

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