

A finite element time domain method for Maxwell's equations

Herbert Egger, **Bogdan Radu**

Johann Radon Institute for Computational and Applied Mathematics (RICAM)
Austrian Academy of Sciences (ÖAW)
Linz, Austria

25th October 2022
RICAM Group seminar



We present **two** discretization methods for solving **Maxwell's equations** on unstructured grids efficiently.

We present **two** discretization methods for solving **Maxwell's equations** on unstructured grids efficiently.

- ▶ The **first method** is based on a finite-element approximation by Nédélec elements of type II, which allows for **mass-lumping by numerical quadrature**. The resulting scheme has **two degrees of freedom per edge** and a block diagonal mass matrix, and hence allows for an efficient time integration by explicit methods.

We present **two** discretization methods for solving **Maxwell's equations** on unstructured grids efficiently.

- ▶ The **first method** is based on a finite-element approximation by Nédélec elements of type II, which allows for **mass-lumping by numerical quadrature**. The resulting scheme has **two degrees of freedom per edge** and a block diagonal mass matrix, and hence allows for an efficient time integration by explicit methods.
- ▶ The **second method** is obtained by an **algebraic reduction of the first**, leading to a scheme with only **one degree of freedom per edge** most of the time. This scheme has a sparse inverse mass matrix, again enabling an efficient time integration

Maxwell's equations

$$\operatorname{curl} \mathbf{H}(t) = \partial_t \mathbf{D}(t) + \mathbf{j}(t) \quad \text{Ampere/Maxwell Law}$$

$$\operatorname{curl} \mathbf{E}(t) = -\partial_t \mathbf{B}(t) \quad \text{Faraday Law}$$

$$\operatorname{div} \mathbf{D}(t) = \varrho \quad \text{Gauss Law}$$

$$\operatorname{div} \mathbf{B}(t) = 0 \quad \text{Magnetic Gauss Law}$$

Material laws for linear and non-dispersive but inhomogeneous/anisotropic media

$$\mathbf{D}(t) = \varepsilon \mathbf{E}(t) \quad \text{and} \quad \mathbf{B}(t) = \mu \mathbf{H}(t) \quad \text{and} \quad \mathbf{j}(t) = \mathbf{j}_s(t) + \sigma \mathbf{E}(t)$$

Substituting the material laws yields

$$\begin{aligned} \varepsilon \partial_t \mathbf{E}(t) + \sigma \mathbf{E}(t) - \operatorname{curl} \mathbf{H}(t) &= -\mathbf{j}_s(t) \\ \mu \partial_t \mathbf{H}(t) + \operatorname{curl} \mathbf{E}(t) &= 0 \end{aligned}$$

Initial and boundary conditions will be discussed in a bit.

$$\begin{aligned}\varepsilon \partial_t \mathbf{E}(t) + \sigma \mathbf{E}(t) - \operatorname{curl} \mathbf{H}(t) &= -\mathbf{j}_s(t) \\ \mu \partial_t \mathbf{H}(t) + \operatorname{curl} \mathbf{E}(t) &= 0\end{aligned}$$

Goal: systematic and flexible space discretization

- ▶ stable: no artificial energy production
- ▶ accurate: provable convergence rates
- ▶ efficient: appropriate for explicit time-stepping methods

$$\begin{aligned}\varepsilon \partial_t \mathbf{E}(t) + \sigma \mathbf{E}(t) - \operatorname{curl} \mathbf{H}(t) &= -\mathbf{j}_s(t) \\ \mu \partial_t \mathbf{H}(t) + \operatorname{curl} \mathbf{E}(t) &= 0\end{aligned}$$

Goal: systematic and flexible space discretization

- ▶ stable: no artificial energy production
- ▶ accurate: provable convergence rates
- ▶ efficient: appropriate for explicit time-stepping methods

Methods: FDTD/FIT, FEM, FVM, DG, ...

- ▶ 1966 - Yee - Numerical solution of initial boundary value problems involving Maxwell's equations in isotropic media
- ▶ 1977 - Weiland - Eine Methode zur Lösung der Maxwell'schen Gleichungen für sechskomponentige Felder auf diskreter Basis
- ▶ 1980 - Taflov - Application of the Finite-Difference Time-Domain method to sinusoidal steady-state electromagnetic penetration problems

$$\begin{aligned}\varepsilon \partial_t \mathbf{E}(t) + \sigma \mathbf{E}(t) - \operatorname{curl} \mathbf{H}(t) &= -\mathbf{j}_s(t) \\ \mu \partial_t \mathbf{H}(t) + \operatorname{curl} \mathbf{E}(t) &= 0\end{aligned}$$

Finite differences : TE case

$$\mathbf{E} = \begin{pmatrix} E_x \\ E_y \\ 0 \end{pmatrix} \quad \mathbf{H} = \begin{pmatrix} 0 \\ 0 \\ H_z \end{pmatrix}$$

$$\begin{cases} \varepsilon \partial_t E_x + \sigma E_x - \partial_y H_z = -j_{s,1}, \\ \varepsilon \partial_t E_y + \sigma E_y + \partial_x H_z = -j_{s,2}, \\ -\mu \partial_t H_z - \partial_x E_y + \partial_y E_x = 0. \end{cases}$$

$$\begin{aligned}\varepsilon \partial_t \mathbf{E}(t) + \sigma \mathbf{E}(t) - \operatorname{curl} \mathbf{H}(t) &= -\mathbf{j}_s(t) \\ \mu \partial_t \mathbf{H}(t) + \operatorname{curl} \mathbf{E}(t) &= 0\end{aligned}$$

Finite differences : TE case

$$\mathbf{E} = \begin{pmatrix} E_x \\ E_y \\ 0 \end{pmatrix} \quad \mathbf{H} = \begin{pmatrix} 0 \\ 0 \\ H_z \end{pmatrix}$$

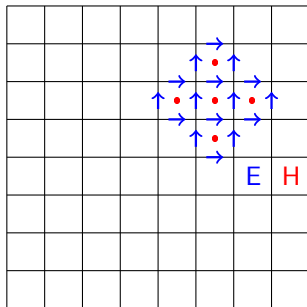
$$\begin{cases} \varepsilon \partial_t E_x + \sigma E_x - \partial_y H_z = -j_{s,1}, \\ \varepsilon \partial_t E_y + \sigma E_y + \partial_x H_z = -j_{s,2}, \\ -\mu \partial_t H_z - \partial_x E_y + \partial_y E_x = 0. \end{cases}$$

$$\begin{aligned}\varepsilon \partial_t \mathbf{E}(t) + \sigma \mathbf{E}(t) - \text{curl } \mathbf{H}(t) &= -\mathbf{j}_s(t) \\ \mu \partial_t \mathbf{H}(t) + \text{curl } \mathbf{E}(t) &= 0\end{aligned}$$

Finite differences : TE case

$$\mathbf{E} = \begin{pmatrix} E_x \\ E_y \\ 0 \end{pmatrix} \quad \mathbf{H} = \begin{pmatrix} 0 \\ 0 \\ H_z \end{pmatrix}$$

$$\begin{cases} \varepsilon \partial_t E_x + \sigma E_x - \partial_y H_z = -j_{s,1}, \\ \varepsilon \partial_t E_y + \sigma E_y + \partial_x H_z = -j_{s,2}, \\ -\mu \partial_t H_z - \partial_x E_y + \partial_y E_x = 0. \end{cases}$$

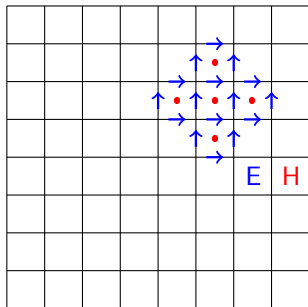


$$\begin{aligned}\varepsilon \partial_t \mathbf{E}(t) + \sigma \mathbf{E}(t) - \text{curl } \mathbf{H}(t) &= -\mathbf{j}_s(t) \\ \mu \partial_t \mathbf{H}(t) + \text{curl } \mathbf{E}(t) &= 0\end{aligned}$$

Finite differences : TE case

$$\mathbf{E} = \begin{pmatrix} E_x \\ E_y \\ 0 \end{pmatrix} \quad \mathbf{H} = \begin{pmatrix} 0 \\ 0 \\ H_z \end{pmatrix}$$

$$\begin{cases} \varepsilon \partial_t E_x + \sigma E_x - \partial_y H_z = -j_{s,1}, \\ \varepsilon \partial_t E_y + \sigma E_y + \partial_x H_z = -j_{s,2}, \\ -\mu \partial_t H_z - \partial_x E_y + \partial_y E_x = 0. \end{cases}$$



Pros

- ▶ Easy to implement
- ▶ stable, $h^2 + \tau^2$ accurate, efficient

Cons

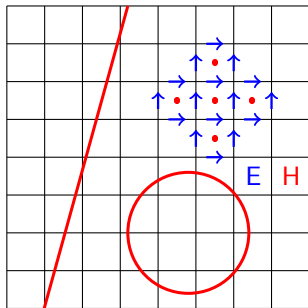
- ▶ Complex domains

$$\begin{aligned}\varepsilon \partial_t \mathbf{E}(t) + \sigma \mathbf{E}(t) - \text{curl } \mathbf{H}(t) &= -\mathbf{j}_s(t) \\ \mu \partial_t \mathbf{H}(t) + \text{curl } \mathbf{E}(t) &= 0\end{aligned}$$

Finite differences : TE case

$$\mathbf{E} = \begin{pmatrix} E_x \\ E_y \\ 0 \end{pmatrix} \quad \mathbf{H} = \begin{pmatrix} 0 \\ 0 \\ H_z \end{pmatrix}$$

$$\begin{cases} \varepsilon \partial_t E_x + \sigma E_x - \partial_y H_z = -j_{s,1}, \\ \varepsilon \partial_t E_y + \sigma E_y + \partial_x H_z = -j_{s,2}, \\ -\mu \partial_t H_z - \partial_x E_y + \partial_y E_x = 0. \end{cases}$$



Pros

- ▶ Easy to implement
- ▶ stable, $h^2 + \tau^2$ accurate, efficient

Cons

- ▶ Complex domains

$$\begin{aligned}\varepsilon \partial_t \mathbf{E}(t) + \sigma \mathbf{E}(t) - \operatorname{curl} \mathbf{H}(t) &= -\mathbf{j}_s(t) \\ \mu \partial_t \mathbf{H}(t) + \operatorname{curl} \mathbf{E}(t) &= 0\end{aligned}$$

$$\begin{aligned}\varepsilon \partial_t \mathbf{E}(t) + \sigma \mathbf{E}(t) - \operatorname{curl} \mathbf{H}(t) &= -\mathbf{j}_s(t) && \text{in } \Omega, t > 0 \\ \mu \partial_t \mathbf{H}(t) + \operatorname{curl} \mathbf{E}(t) &= 0 && \text{in } \Omega, t > 0 \\ \mathbf{n} \times \mathbf{E}(t) &= \mathbf{G}(t) && \text{on } \partial\Omega, t > 0 \\ \mathbf{E}(0) = \mathbf{E}_0, \mathbf{H}(0) &= \mathbf{H}_0 && \text{in } \Omega\end{aligned}$$

$$\begin{aligned}\varepsilon \partial_{tt} \mathbf{E}(t) + \sigma \partial_t \mathbf{E}(t) + \operatorname{curl}(\mu^{-1} \operatorname{curl} \mathbf{E}(t)) &= -\partial_t \mathbf{j}_s(t) && \text{in } \Omega, t > 0 \\ \mathbf{n} \times \mathbf{E}(t) &= \mathbf{G}(t) && \text{on } \partial\Omega, t > 0 \\ \mathbf{E}(0) = \mathbf{E}_0, \partial_t \mathbf{E}(0) &= \mathbf{F}_0 && \text{in } \Omega\end{aligned}$$

$$\begin{aligned}\varepsilon \partial_{tt} \mathbf{E}(t) + \cancel{\sigma} \partial_t \mathbf{E}(t) + \operatorname{curl}(\mu^{-1} \operatorname{curl} \mathbf{E}(t)) &= \cancel{-\partial_t \mathbf{j}_s(t)} && \text{in } \Omega, t > 0 \\ \mathbf{n} \times \mathbf{E}(t) &= \cancel{\mathbf{G}(t)} && \text{on } \partial\Omega, t > 0 \\ \mathbf{E}(0) = \mathbf{E}_0, \partial_t \mathbf{E}(0) &= \mathbf{F}_0 && \text{in } \Omega\end{aligned}$$

$$\begin{aligned}\varepsilon \partial_{tt} \mathbf{E}(t) + \cancel{\sigma} \partial_t \mathbf{E}(t) + \operatorname{curl}(\cancel{\mu}^{-1} \operatorname{curl} \mathbf{E}(t)) &= \cancel{-\partial_t \mathbf{j}_s(t)} && \text{in } \Omega, t > 0 \\ \mathbf{n} \times \mathbf{E}(t) &= \cancel{\mathbf{G}(t)} && \text{on } \partial\Omega, t > 0 \\ \mathbf{E}(0) = \mathbf{E}_0, \partial_t \mathbf{E}(0) &= \mathbf{F}_0 && \text{in } \Omega\end{aligned}$$

$$\begin{aligned}\varepsilon \partial_{tt} \mathbf{E}(t) + \operatorname{curl}(\operatorname{curl} \mathbf{E}(t)) &= 0 && \text{in } \Omega, t > 0 \\ \mathbf{n} \times \mathbf{E}(t) &= 0 && \text{on } \partial\Omega, t > 0 \\ \mathbf{E}(0) = \mathbf{E}_0, \partial_t \mathbf{E}(0) &= \mathbf{F}_0 && \text{in } \Omega\end{aligned}$$

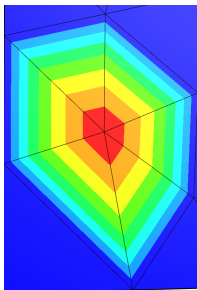
Galerkin method: For $t > 0$ find $\mathbf{E}_h(t) \in \mathbf{V}_h \subseteq H(\operatorname{curl}, \Omega)$ such that

$$(\varepsilon \partial_{tt} \mathbf{E}_h(t), \mathbf{v}_h) + (\operatorname{curl} \mathbf{E}_h(t), \operatorname{curl} \mathbf{v}_h) = 0$$

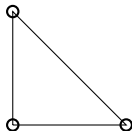
for all test functions $\mathbf{v}_h \in \mathbf{V}_h$.

Finite element spaces: What are proper spaces for discretizing $H(\operatorname{curl}, \Omega)$?

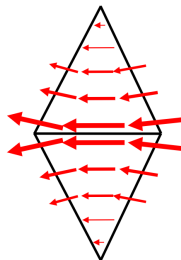
- ▶ Global basis functions for H^1



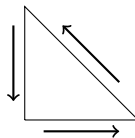
- ▶ Local degrees of freedom



- ▶ Global basis functions for $H(\text{curl})$

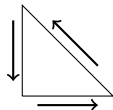


- ▶ Local degrees of freedom



Finite element spaces on reference elements.

► 1980 - Nedelec - Mixed Finite Elements in \mathbb{R}^3

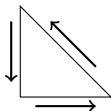


$$V_h(T) = \mathcal{N}_0(T)$$

$$\begin{aligned}\phi_1 &= (1 - y, x) & \phi_3 &= (y, 1 - x) \\ \phi_2 &= (-y, x)\end{aligned}$$

Finite element spaces on reference elements.

- ▶ 1980 - Nedelec - Mixed Finite Elements in \mathbb{R}^3



$$V_h(T) = \mathcal{N}_0(T) \quad \begin{aligned} \phi_1 &= (1 - y, x) & \phi_3 &= (y, 1 - x) \\ \phi_2 &= (-y, x) \end{aligned}$$

Lemma (accuracy) If \mathbf{E} is sufficiently smooth, then

$$\|\mathbf{E}(t) - \mathbf{E}_h(t)\|_{H(\text{curl})} \leq Ch$$

- ▶ 1992 - Monk - Analysis of a finite element method for Maxwell's equations
- ▶ 1993 - Monk - An analysis of Nedelec's method for spatial discretization of Maxwell's equations

Stability and accuracy.

Lowest order MFEM yields stable and accurate approximation in space.

Stability and accuracy.

Lowest order MFEM yields stable and accurate approximation in space.

Numerical solution. Time integration of resulting ode system

$$M_\varepsilon \partial_{tt} \mathbf{e}(t) + K \mathbf{e}(t) = 0$$

by explicit schemes requires application of M_ε^{-1} .

Stability and accuracy.

Lowest order MFEM yields stable and accurate approximation in space.

Numerical solution. Time integration of resulting ode system

$$\mathbf{M}_\varepsilon \partial_{tt} \mathbf{e}(t) + \mathbf{K} \mathbf{e}(t) = 0$$

by explicit schemes requires application of $\mathbf{M}_\varepsilon^{-1}$.

Note. \mathbf{M}_ε does not have a sparse inverse!

Thus, explicit time-stepping for standard FEM is not efficient.

Stability and accuracy.

Lowest order MFEM yields stable and accurate approximation in space.

Numerical solution. Time integration of resulting ode system

$$M_\varepsilon \partial_{tt} \mathbf{e}(t) + K \mathbf{e}(t) = 0$$

by explicit schemes requires application of M_ε^{-1} .

Note. M_ε does not have a sparse inverse!

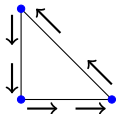
Thus, explicit time-stepping for standard FEM is not efficient.

Remedy – Mass-lumping: replace M_ε by approximation M_ε^L such that

- ▶ M_ε^L corresponds to positive definite matrix (stability)
- ▶ M_ε^L is good approximation for M_ε (accuracy)
- ▶ $(M_\varepsilon^L)^{-1}$ can be applied efficiently (efficiency)

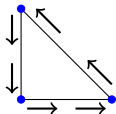
construction of M_ε^L usually via numerical quadrature.

Use a larger polynomial space



$$\tilde{\mathbf{V}}_h(T) = \mathcal{NC}_1(T) = P_1(T)^2$$

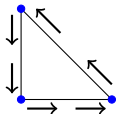
Use a larger polynomial space



$$\tilde{\mathbf{V}}_h(T) = \mathcal{NC}_1(T) = P_1(T)^2$$

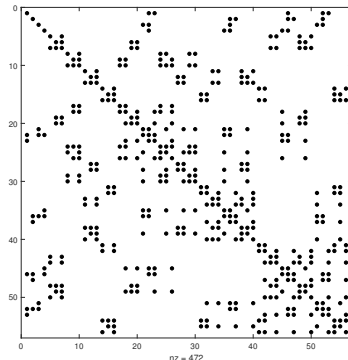
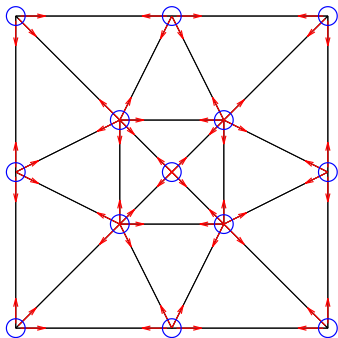
Lemma. $\tilde{\mathbf{M}}_\varepsilon^L$ is block diagonal and thus also $(\tilde{\mathbf{M}}_\varepsilon^L)^{-1}$.

Use a larger polynomial space

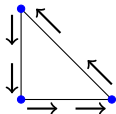


$$\tilde{V}_h(T) = \mathcal{NC}_1(T) = P_1(T)^2$$

Lemma. \tilde{M}_ε^L is block diagonal and thus also $(\tilde{M}_\varepsilon^L)^{-1}$.

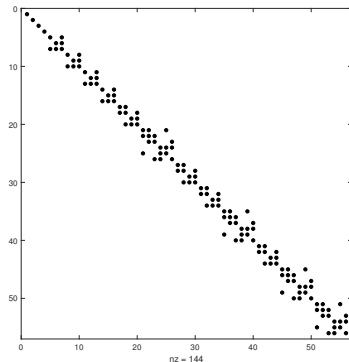
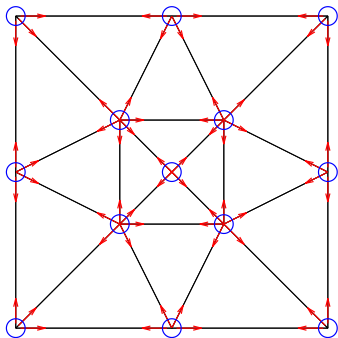


Use a larger polynomial space

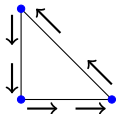


$$\tilde{\mathbf{V}}_h(T) = \mathcal{NC}_1(T) = P_1(T)^2$$

Lemma. $\tilde{\mathbf{M}}_\varepsilon^L$ is block diagonal and thus also $(\tilde{\mathbf{M}}_\varepsilon^L)^{-1}$.

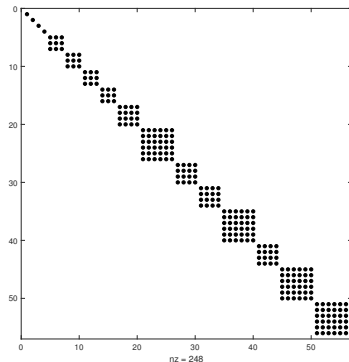
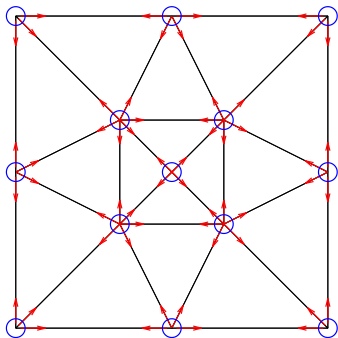


Use a larger polynomial space

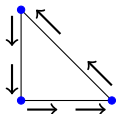


$$\tilde{V}_h(T) = \mathcal{NC}_1(T) = P_1(T)^2$$

Lemma. \tilde{M}_ε^L is block diagonal and thus also $(\tilde{M}_\varepsilon^L)^{-1}$.



Use a larger polynomial space



$$\tilde{\mathbf{V}}_h(T) = \mathcal{NC}_1(T) = P_1(T)^2$$

Lemma. $\tilde{\mathbf{M}}_\varepsilon^L$ is block diagonal and thus also $(\tilde{\mathbf{M}}_\varepsilon^L)^{-1}$.

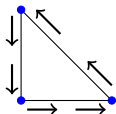
Theorem (accuracy)

If \mathbf{E} is sufficiently smooth, then

$$\|\mathbf{E}(t) - \tilde{\mathbf{E}}_h(t)\|_{H(\text{curl})} \leq Ch$$

- 2020 - Egger, Radu - A mass-lumped mixed finite element method for Maxwell's equations

Use a larger polynomial space



$$\tilde{\mathbf{V}}_h(T) = \mathcal{NC}_1(T) = P_1(T)^2$$

Lemma. $\tilde{\mathbf{M}}_\varepsilon^L$ is block diagonal and thus also $(\tilde{\mathbf{M}}_\varepsilon^L)^{-1}$.

Theorem (accuracy)

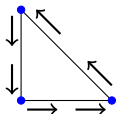
If \mathbf{E} is sufficiently smooth, then

$$\|\mathbf{E}(t) - \tilde{\mathbf{E}}_h(t)\|_{H(\text{curl})} \leq Ch$$

- 2020 - Egger, Radu - A mass-lumped mixed finite element method for Maxwell's equations

Proof Idea: Error splitting in discrete and projection error, discrete stability, energy estimates, consistency error, **analysis of the quadrature error (Strang)**.

Use a larger polynomial space



$$\tilde{\mathbf{V}}_h(T) = \mathcal{NC}_1(T) = P_1(T)^2$$

Lemma. $\tilde{\mathbf{M}}_\varepsilon^L$ is block diagonal and thus also $(\tilde{\mathbf{M}}_\varepsilon^L)^{-1}$.

Theorem (accuracy)

If \mathbf{E} is sufficiently smooth, then

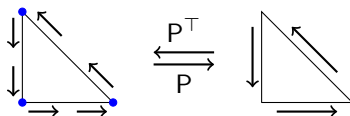
$$\|\mathbf{E}(t) - \tilde{\mathbf{E}}_h(t)\|_{H(\text{curl})} \leq Ch$$

- 2020 - Egger, Radu - A mass-lumped mixed finite element method for Maxwell's equations

Proof Idea: Error splitting in discrete and projection error, discrete stability, energy estimates, consistency error, **analysis of the quadrature error (Strang)**.

Requirement : The quadrature rule must be exact for $P_0(T)^2 \times \tilde{\mathbf{V}}_h(T)$

Consider the projection $\Pi_P : \mathcal{NC}_1 \rightarrow \mathcal{N}_0$ realized algebraically by the matrix P :



such that $PP^T = \text{Id}$. Then

$$M := P\tilde{M}P^T \quad \text{and} \quad K := P\tilde{K}P^T$$

are mass and stiffness matrices for \mathcal{N}_0 w.r.t. the projected basis. Moreover,

$$\tilde{K} = P^TKP$$

since both \mathcal{NC}_1 and \mathcal{N}_0 produce the same curls, i.e. $\text{curl} \Pi_P v_h = \text{curl} v_h$.

This is a consequence of the **commuting diagram property**

Let $\tilde{\mathbf{e}}(t)$ be the coefficient vector to $\mathbf{E}_h(t)$, and it satisfies

$$\tilde{\mathbf{M}}_\varepsilon^L \partial_{tt} \tilde{\mathbf{e}}(t) + \tilde{\mathbf{K}} \tilde{\mathbf{e}}(t) = 0$$

Then $\mathbf{e}(t) := \mathbf{P} \tilde{\mathbf{e}}(t)$ satisfies

$$\begin{aligned} \partial_{tt} \mathbf{e}(t) &= \mathbf{P} \partial_{tt} \tilde{\mathbf{e}}(t) = -\mathbf{P} (\tilde{\mathbf{M}}_\varepsilon^L)^{-1} \tilde{\mathbf{K}} \tilde{\mathbf{e}}(t) = -\mathbf{P} (\tilde{\mathbf{M}}_\varepsilon^L)^{-1} \mathbf{P}^\top \mathbf{K} \mathbf{P} \tilde{\mathbf{e}}(t) \\ &= -\mathbf{P} (\tilde{\mathbf{M}}_\varepsilon^L)^{-1} \mathbf{P}^\top \mathbf{K} \mathbf{e}(t) \end{aligned}$$

Thus $\mathbf{e}(t)$ satisfies

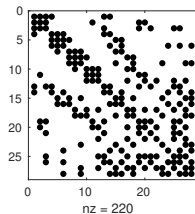
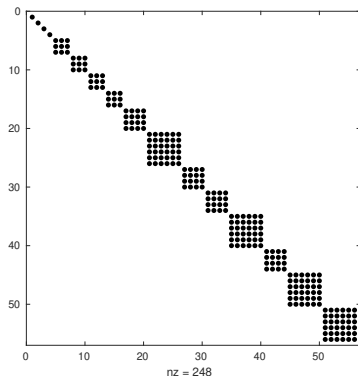
$$\mathbf{M}_\varepsilon^L \partial_{tt} \mathbf{e}(t) + \mathbf{K} \mathbf{e}(t) = 0$$

where $(\mathbf{M}_\varepsilon^L)^{-1} = \mathbf{P} (\tilde{\mathbf{M}}_\varepsilon^L)^{-1} \mathbf{P}^\top$. **Note:** \mathbf{M}_ε^L is not a FEM matrix.

Theorem (accuracy) If \mathbf{E} is sufficiently smooth, then $\mathbf{E}_h(t) := \Pi_{\mathbf{P}} \tilde{\mathbf{E}}_h(t)$ satisfies

$$\|\mathbf{E}(t) - \mathbf{E}_h(t)\|_{H(\text{curl})} \leq Ch$$

Let us investigate the impact of the reduction (visually)



Inverse of \tilde{M}_ε on the left, the inverse of M_ε on the right.

$$(\varepsilon \partial_{tt} \tilde{\mathbf{E}}_h(t), \tilde{\mathbf{v}}_h)_h + (\operatorname{curl} \tilde{\mathbf{E}}_h(t), \operatorname{curl} \tilde{\mathbf{v}}_h) = 0$$

$$\Downarrow$$

$$\mathbf{e}(t) := \mathbf{P} \tilde{\mathbf{e}}(t)$$

$$\Downarrow$$

$$\partial_{tt} \mathbf{e}(t) = -(\mathbf{P}(\tilde{\mathbf{M}}_\varepsilon^L)^{-1} \mathbf{P}^\top) \mathbf{K} \mathbf{e}(t)$$

$$(\varepsilon d_{\tau\tau} \tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\operatorname{curl} \tilde{\mathbf{E}}_h^n, \operatorname{curl} \tilde{\mathbf{v}}_h) = 0$$

$$\Downarrow$$

$$\mathbf{e}^n := \mathbf{P} \tilde{\mathbf{e}}^n$$

$$\Downarrow$$

$$\frac{\mathbf{e}^{n+1} - 2\mathbf{e}^n + \mathbf{e}^{n-1}}{\tau^2} = -(\mathbf{P}(\tilde{\mathbf{M}}_\varepsilon^L)^{-1}\mathbf{P}^\top)\mathbf{K}\mathbf{e}^n$$

$$(\varepsilon d_{\tau\tau} \tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\operatorname{curl} \tilde{\mathbf{E}}_h^n, \operatorname{curl} \tilde{\mathbf{v}}_h) = 0$$

$$\Downarrow$$

$$\mathbf{e}^n := \mathbf{P} \tilde{\mathbf{e}}^n$$

$$\Downarrow$$

$$\mathbf{e}^{n+1} = 2\mathbf{e}^n - \mathbf{e}^{n-1} - \tau^2 (\mathbf{P}(\tilde{\mathbf{M}}_\varepsilon^L)^{-1} \mathbf{P}^\top) \mathbf{K} \mathbf{e}^n$$

$$(\varepsilon d_{\tau\tau} \tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\sigma d_{\tau} \tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\operatorname{curl} \tilde{\mathbf{E}}_h^n, \operatorname{curl} \tilde{\mathbf{v}}_h) = 0$$

$$\Downarrow$$

$$\mathbf{e}^n := \mathbf{P} \tilde{\mathbf{e}}^n$$

$$\Downarrow$$

$$\begin{aligned} \mathbf{e}^{n+1} = & 2\mathbf{e}^n - \mathbf{e}^{n-1} - \tau^2 (\mathbf{P}(\tilde{\mathbf{M}}_{\varepsilon}^L + \frac{\tau}{2} \tilde{\mathbf{M}}_{\sigma}^L)^{-1} \mathbf{P}^{\top}) \mathbf{K} \mathbf{e}^n \\ & - \tau^2 (\mathbf{P}(\tilde{\mathbf{M}}_{\varepsilon}^L + \frac{\tau}{2} \tilde{\mathbf{M}}_{\sigma}^L)^{-1} \left(\tilde{\mathbf{M}}_{\sigma}^L \frac{\tilde{\mathbf{e}}^n - \tilde{\mathbf{e}}^{n-1}}{\tau} \right) \end{aligned}$$

$$(\varepsilon d_{\tau\tau} \tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h) + (\sigma d_{\tau} \tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\operatorname{curl} \tilde{\mathbf{E}}_h^n, \operatorname{curl} \tilde{\mathbf{v}}_h) = 0$$

$$\begin{array}{c} \text{\textcolor{red}{X}} \\ \mathbf{e}^n := \mathbf{P} \tilde{\mathbf{e}}^n \\ \text{\textcolor{red}{X}} \end{array}$$

$$\begin{aligned} \mathbf{e}^{n+1} &:= 2\mathbf{e}^n - \mathbf{e}^{n-1} - \tau^2 (\mathbf{P}(\tilde{\mathbf{M}}_{\varepsilon}^L + \frac{\tau}{2} \tilde{\mathbf{M}}_{\sigma}^L)^{-1} \mathbf{P}^{\top}) \mathbf{K} \mathbf{e}^n \\ &\quad - \tau^2 (\mathbf{P}(\tilde{\mathbf{M}}_{\varepsilon}^L + \frac{\tau}{2} \tilde{\mathbf{M}}_{\sigma}^L)^{-1} \left(\text{\textcolor{red}{P}}^{\top} \text{\textcolor{red}{M}}_{\sigma} \text{\textcolor{red}{P}} \frac{\tilde{\mathbf{e}}^n - \tilde{\mathbf{e}}^{n-1}}{\tau} \right) \end{aligned}$$

Recall:

$$\tilde{\mathbf{K}} = \mathbf{P}^{\top} \mathbf{K} \mathbf{P} \quad \text{but} \quad \tilde{\mathbf{M}}_{\sigma}^L \neq \text{\textcolor{red}{P}}^{\top} \text{\textcolor{red}{M}}_{\sigma} \text{\textcolor{red}{P}}$$

$$(\varepsilon d_{\tau\tau} \tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\sigma d_{\tau} \tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\operatorname{curl} \tilde{\mathbf{E}}_h^n, \operatorname{curl} \tilde{\mathbf{v}}_h) = 0$$

$$\begin{array}{c} \text{\textcolor{red}{X}} \\ \mathbf{e}^n := \mathbf{P} \tilde{\mathbf{e}}^n \\ \text{\textcolor{red}{X}} \end{array}$$

$$\begin{aligned} \mathbf{e}^{n+1} & \textcolor{red}{:=} 2\mathbf{e}^n - \mathbf{e}^{n-1} - \tau^2 (\mathbf{P}(\tilde{\mathbf{M}}_{\varepsilon}^L + \frac{\tau}{2} \tilde{\mathbf{M}}_{\sigma}^L)^{-1} \mathbf{P}^{\top}) \mathbf{K} \mathbf{e}^n \\ & - \tau^2 (\mathbf{P}(\tilde{\mathbf{M}}_{\varepsilon}^L + \frac{\tau}{2} \tilde{\mathbf{M}}_{\sigma}^L)^{-1} \mathbf{P}^{\top} \mathbf{M}_{\sigma} \frac{\mathbf{e}^n - \mathbf{e}^{n-1}}{\tau} \end{aligned}$$

$$((\varepsilon + \frac{\tau}{2}\sigma)d_{\tau\tau}\tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\sigma\Pi_P d_\tau \tilde{\mathbf{E}}_h^{n-1/2}, \Pi_P \tilde{\mathbf{v}}_h) + (\operatorname{curl} \tilde{\mathbf{E}}_h^n, \operatorname{curl} \tilde{\mathbf{v}}_h) = 0$$

$$\Downarrow$$

$$\mathbf{e}^n := P\tilde{\mathbf{e}}^n$$

$$\Downarrow$$

$$\begin{aligned} \mathbf{e}^{n+1} &:= 2\mathbf{e}^n - \mathbf{e}^{n-1} - \tau^2(P(\tilde{M}_\varepsilon^L + \frac{\tau}{2}\tilde{M}_\sigma^L)^{-1}P^\top)K\mathbf{e}^n \\ &\quad - \tau^2(P(\tilde{M}_\varepsilon^L + \frac{\tau}{2}\tilde{M}_\sigma^L)^{-1}P^\top M_\sigma \frac{\mathbf{e}^n - \mathbf{e}^{n-1}}{\tau} \end{aligned}$$

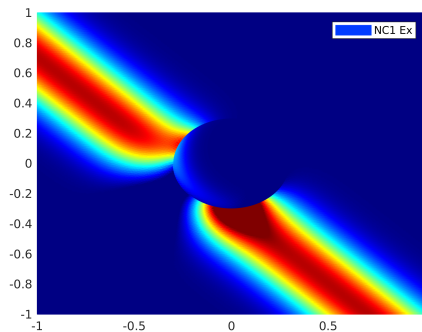


Figure: \mathcal{NC}_1 method with mass-lumping

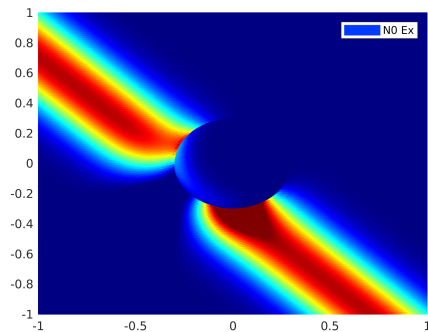
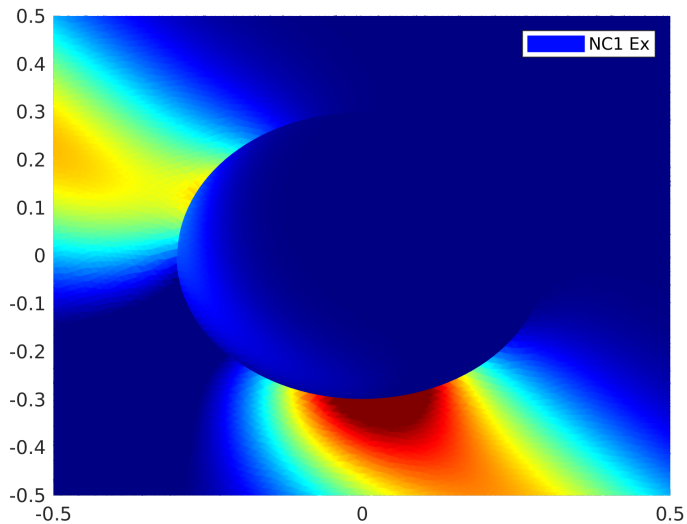
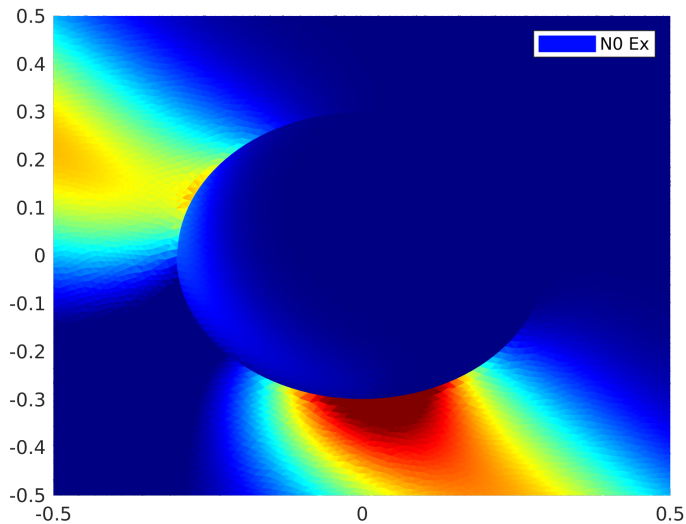


Figure: Reduced \mathcal{NC}_1 method (to \mathcal{N}_0)





$$((\varepsilon + \frac{\tau}{2}\sigma)d_{\tau\tau}\tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\sigma\Pi_P d_\tau \tilde{\mathbf{E}}_h^{n-1/2}, \Pi_P \tilde{\mathbf{v}}_h) + (\operatorname{curl} \tilde{\mathbf{E}}_h^n, \operatorname{curl} \tilde{\mathbf{v}}_h) = 0$$

Theorem (accuracy) If \mathbf{E} is sufficiently smooth, then $\mathbf{E}_h^n := \Pi_P \tilde{\mathbf{E}}_h^n$ satisfies

$$\|\mathbf{E}^n - \mathbf{E}_h^n\|_{H(\operatorname{curl})} \leq C(h^{1/2} + \tau^2)$$

$$((\varepsilon + \frac{\tau}{2}\sigma)d_{\tau\tau}\tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\sigma\Pi_P d_{\tau}\tilde{\mathbf{E}}_h^{n-1/2}, \Pi_P\tilde{\mathbf{v}}_h) + (\operatorname{curl}\tilde{\mathbf{E}}_h^n, \operatorname{curl}\tilde{\mathbf{v}}_h) = 0$$

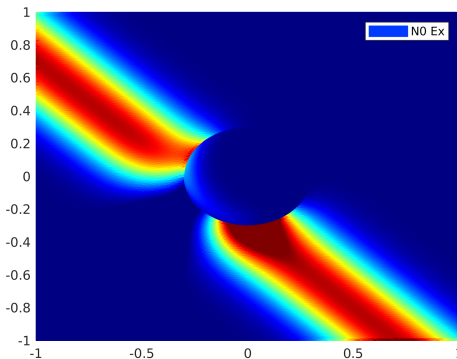
Theorem (accuracy) If \mathbf{E} is sufficiently smooth, then $\mathbf{E}_h^n := \Pi_P\tilde{\mathbf{E}}_h^n$ satisfies

$$\|\mathbf{E}^n - \mathbf{E}_h^n\|_{H(\operatorname{curl})} \leq C(h^{1/2} + \tau^2)$$

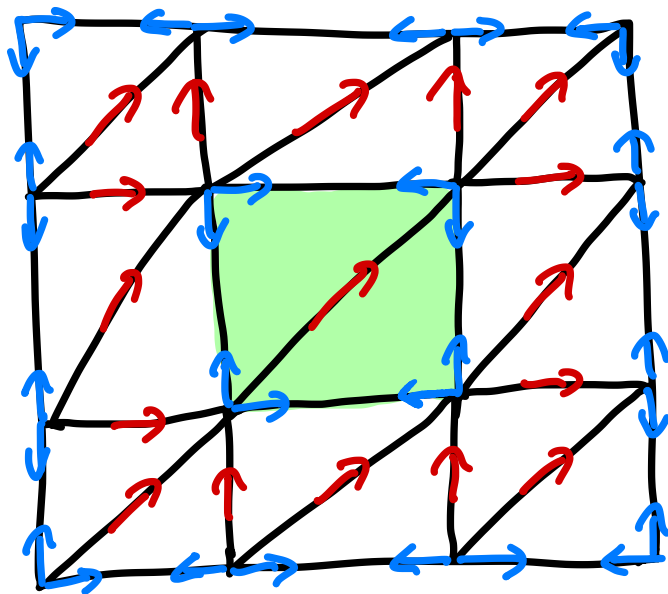
Further complications for non-trivial boundary conditions

$\mathbf{n} \times \mathbf{E}(t) = \mathbf{G}(t)$. Again:

$$\|\mathbf{E}^n - \mathbf{E}_h^n\|_{H(\operatorname{curl})} \leq C(h^{1/2} + \tau^2)$$



Let's fix this!



$$\sigma \neq 0$$

$$\mathcal{N}_0^\dagger := \text{span} \left\{ \begin{array}{ll} \Phi_0^e & \text{if } e \text{ is an interior edge and } \sigma \text{ is smooth across it} \\ \Phi_1^{e,1}, \Phi_1^{e,2} & \text{if } e \text{ is a boundary edge and } \sigma \text{ jumps across it} \end{array} \right\}$$

Define the projection $\Pi_P^\dagger : \mathcal{NC}_1 \rightarrow \mathcal{N}_0^\dagger$ and the method:

$$((\varepsilon + \frac{\tau}{2}\sigma)d_{\tau\tau}\tilde{\mathbf{E}}_h^n, \tilde{\mathbf{v}}_h)_h + (\sigma\Pi_P^\dagger d_\tau\tilde{\mathbf{E}}_h^{n-1/2}, \Pi_P^\dagger \tilde{\mathbf{v}}_h) + (\text{curl } \tilde{\mathbf{E}}_h^n, \text{curl } \tilde{\mathbf{v}}_h) = 0$$

Theorem (accuracy) If \mathbf{E} is sufficiently smooth, then $\mathbf{E}_h^n := \Pi_P^\dagger \tilde{\mathbf{E}}_h^n$ satisfies

$$\|\mathbf{E}^n - \mathbf{E}_h^n\|_{H(\text{curl})} \leq C(h + \tau^2)$$

- ▶ Derived 2 methods