

Adaptive methods for photonic crystal fibers

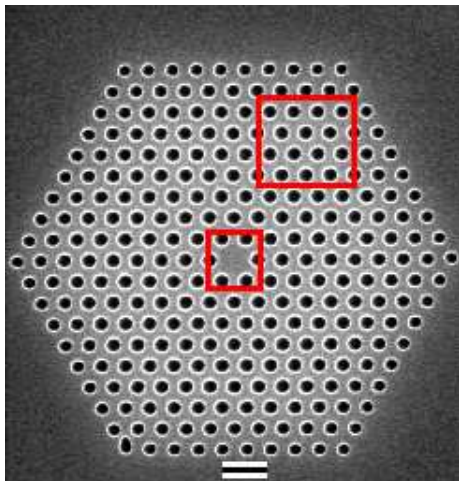
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What is a Photonic Crystal Fiber?



It can be possible for light of certain frequencies to be expelled for all directions - light of these frequencies simply will not propagate in the material. This situation is called a "photonic band gap". A periodic structure could be used to keep light of certain frequencies in the center of the fiber.



Maxwell's Equations

$$\begin{cases} \nabla \times \mathbf{E} = -\frac{i\omega}{c} \mu \mathbf{H}, \\ \nabla \times \mathbf{H} = \frac{i\omega}{c} \varepsilon \mathbf{E}, \\ \nabla \cdot \mu \mathbf{H} = 0, \\ \nabla \cdot \varepsilon \mathbf{E} = 0. \end{cases}$$

$$\begin{aligned} \mu &= 1, \\ \varepsilon &= \begin{pmatrix} \epsilon_{11} & \epsilon_{12} & 0 \\ \epsilon_{21} & \epsilon_{22} & 0 \\ 0 & 0 & \epsilon_{33} \end{pmatrix}, \\ \epsilon_{12} &= \epsilon_{21} \end{aligned}$$

TM Case:

$$\mathbf{E} = (0, 0, u(x, y))$$

$$\mathbf{E} = (u_y(x, y), -u_x(x, y), 0)$$

TE Case:

$$\mathbf{H} = (0, 0, u(x, y))$$

$$\mathbf{E} = (u_y(x, y), -u_x(x, y), 0)$$



Floquet transform

$$g \in L^2(\mathbb{R}^2) \quad (\mathcal{F}g)(\alpha, \mathbf{x}) = e^{-i\alpha \cdot \mathbf{x}} \sum_{\mathbf{n} \in \mathbb{Z}^2} g(\mathbf{x} - \mathbf{n}) e^{i\alpha \cdot \mathbf{n}}$$

$$\alpha \in [-\pi, \pi]^2, \mathbf{x} \in [0, 1]^2$$

⇒ Now the domain of x is a torus and α is a parameter.



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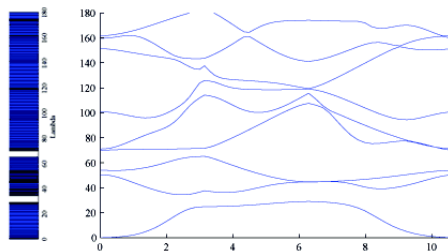


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$$\text{Lemma: } \mathcal{F}(\nabla g) = (\nabla + i\alpha)\mathcal{F}g.$$

$$\text{TE: } \nabla \cdot (M\nabla u) + \lambda u = 0, \quad x \in \mathbb{T}, \alpha \in [-\pi, \pi]^2.$$



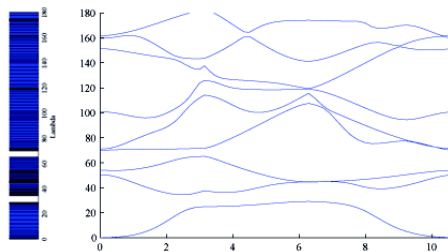


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Lemma: $\mathcal{F}(\nabla g) = (\nabla + i\alpha)\mathcal{F}g.$

TE: $(\nabla + i\alpha) \cdot M(\nabla + i\alpha)u_\alpha + \lambda_\alpha u_\alpha = 0, \quad \mathbf{x} \in \mathbb{T}, \alpha \in [-\pi, \pi]^2.$



Why a Posteriori Error Estimator for TE case?

Reasons:

- Optimize Computational Power
- Regularity in each Subdomain is $H^{1+1/2+\beta}(\mathbb{T})$, $\beta > 0$
- Solutions could oscillate a lot.

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Posteriori Error Estimator

$$a_\alpha(u, v) = \lambda(u, v), \quad (\text{Continuous})$$

$$a_\alpha(u_h, v_h) = \lambda_h(u_h, v_h), \quad (\text{Discrete})$$



- \mathcal{T}_h (Mesh resolves the interface), $V_h \subset H^1(\mathbb{T}, \mathbb{C})$ (Finite Element Space)

- $\eta(\lambda_h, u_h) = \left(\sum_{\tau \in \mathcal{T}_h} \eta_\tau^2(\lambda_h, u_h) \right)^{1/2}$ (Error Estimator)

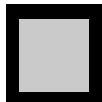
Properties:

1. $\|e\|_1 \leq C\eta + h.o.t.$ (Reliability)
2. $\lambda_h - \lambda \leq C\eta^2 + h.o.t.$ (Reliability)
3. $\eta \leq C\|e_h\|_1 + h.o.t.$ (Efficiency)

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Residuals

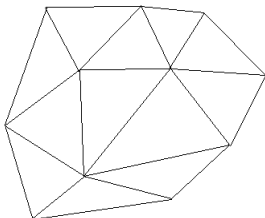
(with isotropic dielectricity ε)

$$\eta = \left(\sum_{\tau \in \mathfrak{T}_h} \eta_\tau^2(\lambda_h, u_h) \right)^{1/2},$$

$$R_I(u_h, \lambda_h)(x) := ((\nabla + i\alpha) \cdot \varepsilon^{-1}(\nabla + i\alpha)u_h + \lambda_h u_h)(x),$$

$$R_E(u_h)(x) := ([\hat{n} \cdot \varepsilon^{-1} \nabla u_h])(x)$$

$$+([\hat{n} \cdot \varepsilon^{-1} i\alpha u_h])(x).$$



Residuals

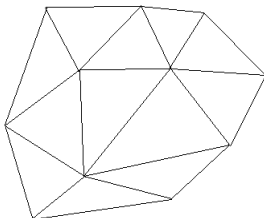
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$$\eta = \left(\sum_{\tau \in \mathfrak{T}_h} h_\tau^2 \varepsilon_\tau \|R_I(u_h, \lambda_h)\|_{0,\tau}^2 + \sum_{e \in \mathfrak{E}_h} h_e \varepsilon_e \|R_E(u_h)\|_{0,e}^2 \right)^{1/2}.$$

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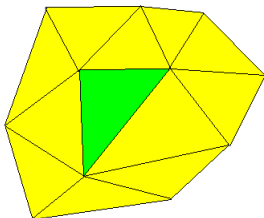
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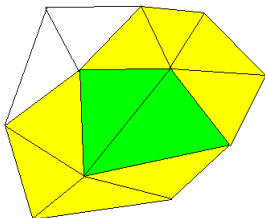
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Shift

Proposition 1:(Gårding's inequality)

For any constant $K \geq \varepsilon_{max} \left(\frac{1}{2} + |\alpha|^2 \right)$, we have

$$a_\alpha(u, u) + K \|u\|_{0,T}^2 \geq \frac{1}{2} \|u\|_{1,\varepsilon,T}^2 \quad \forall u \in H^1(T, \mathbb{C}).$$

Define:

$$\mu \equiv \lambda - K$$

$$\mu_h \equiv \lambda_h - K$$

Non-linearity

LEMMA 2:

$$(\mu u - \mu_h u_h, \mathbf{e})_{0, \mathbb{T}} = \frac{\lambda + \lambda_h + 2K}{2} (\mathbf{e}, \mathbf{e})_{0, \mathbb{T}} + i(\lambda_h - \lambda) \text{Im}(u, u_h)_{0, \mathbb{T}}.$$

LEMMA 3:

$$\begin{aligned} a_\alpha(\mathbf{e}, \mathbf{v}) + K(\mathbf{e}, \mathbf{v})_{0, \mathbb{T}} - (\lambda u - \lambda_h u_h, \mathbf{v})_{0, \mathbb{T}} &= \sum_{\tau \in \mathfrak{T}_h} \int_{\tau} R_I(u_h, \lambda_h) \bar{\mathbf{v}} \, dt \\ &\quad - \sum_{e \in \mathfrak{E}_h} \int_e R_E(u_h) \bar{\mathbf{v}} \, d\gamma. \end{aligned}$$

See [6].

Reliability

THEOREM 4:

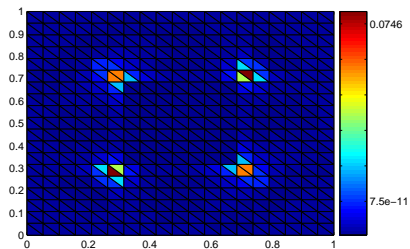
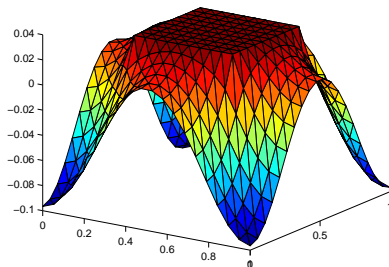
$$\sqrt{a_\alpha(\mathbf{e}, \mathbf{e})} \leq C_{r1}\eta + \frac{\lambda + \lambda_h + 2K}{2} \frac{(e, e)_{0,T}}{\sqrt{a_\alpha(\mathbf{e}, \mathbf{e})}}.$$

THEOREM 5:

$$\begin{aligned} \lambda_h - \lambda &\leq C_{r1}^2\eta^2 + C_{r1}\eta \frac{\lambda_h + \lambda + 2K}{2} \frac{(e, e)_{0,T}}{\sqrt{a_\alpha(\mathbf{e}, \mathbf{e})}} \\ &\quad + \frac{\lambda_h - \lambda}{2} (e, e)_{0,T}. \end{aligned}$$

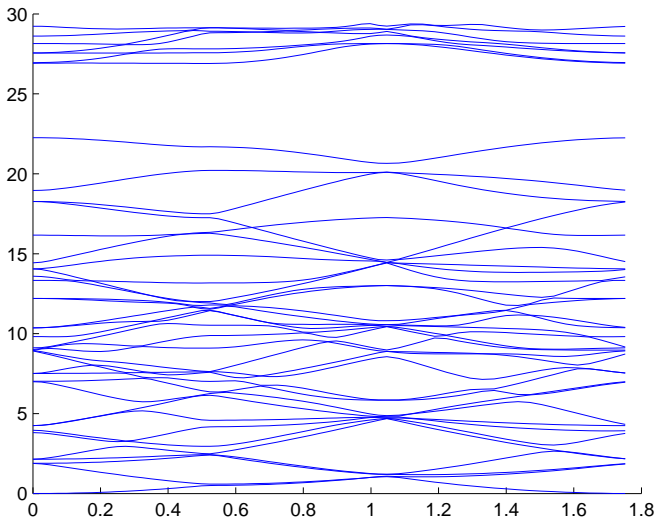
Numerical Results

Eig: 50.4545

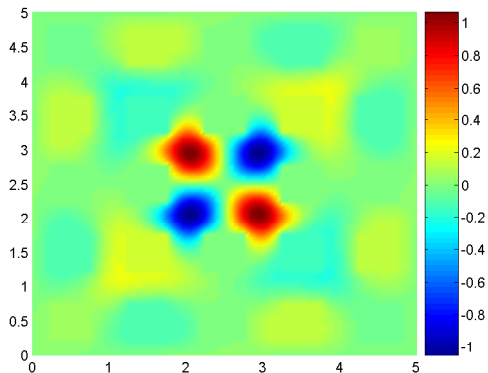
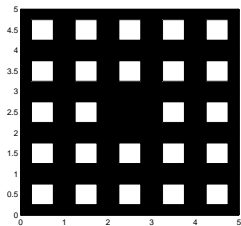




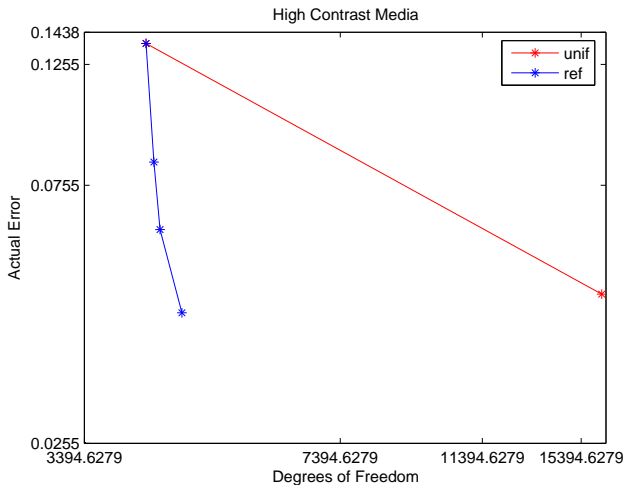
Supercell with defect



Trapped mode



Convergence



References I



I. Babuška and J. Osborn.

Eigenvalue Problems.

Elsevier, 1991.



R. Verfurth.

A review of posteriori error estimation and adaptive mesh refinement technique.

Wiley-Teubner, 1996.



D. C. Dobson.

An Efficient Method for Band Structure Calculations in 2D Photonic Crystal.

JCP, 149:363-376, 1999.

References II



M. Petzoldt.

Regularity and error estimations for elliptic problems with discontinuous coefficients.

2001.



S. Soussi.

Convergence of the supercell method for defect modes calculations in photonic crystals.

2005.



T.F.Walsh , G.M. Reese and U.L. Hetmaniuk.

Explicit a posteriori error estimates for eigenvalue analysis of heterogeneous elastic structures.

Sandia National Laboratories, 2005