

# Using Complex Shifts in Rayleigh Quotient Iteration to Compute Close Eigenvalues

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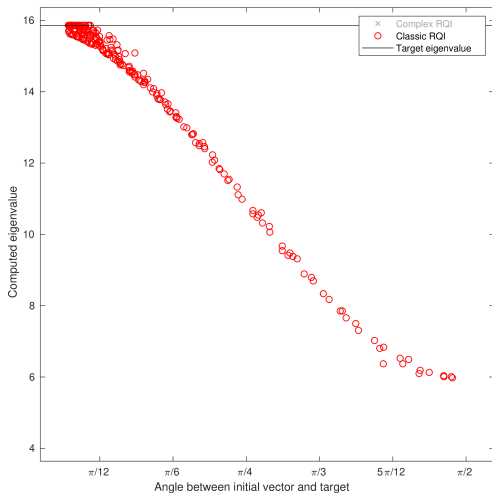
17th June 2020

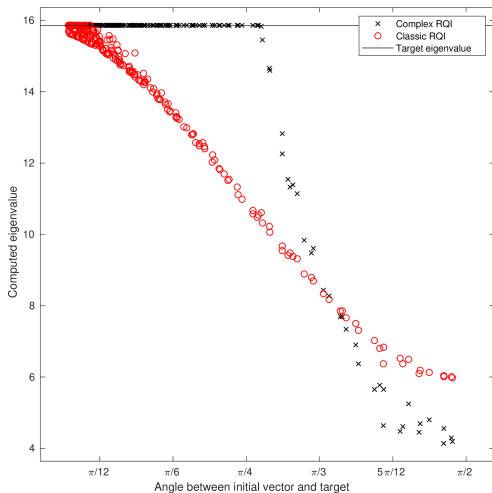


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# Contents

Using Complex Shifts  
in Rayleigh Quotient Iteration  
to Compute Close Eigenvalues





Compute one eigenpair  $(\lambda, \mathbf{v})$  of symmetric matrix  $\mathbf{A}$ , i. e.,

$$\mathbf{A}\mathbf{v} = \lambda\mathbf{v}.$$

### Theorem

- ▶ *The eigenvalues  $\lambda_1, \dots, \lambda_n$  of  $\mathbf{A}$  are real.*
- ▶ *The eigenvectors  $\mathbf{v}_1, \dots, \mathbf{v}_n$  of  $\mathbf{A}$  are real.*
- ▶ *The eigenvectors form an orthogonal basis of  $\mathbb{R}^n$ .*

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**Algorithm:** Power method

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**Input:** Nonzero unit vector  $\mathbf{x}^{(0)}$

**for**  $k = 0, 1, \dots$  **until convergence do**

$\mathbf{x}^{(k+1)} \leftarrow \mathbf{A}\mathbf{x}^{(k)}$

    Normalise  $\mathbf{x}^{(k+1)}$

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## Theorem

- ▶ The eigenvalues of  $A^{-1}$  are  $\frac{1}{\lambda_i}$ .
- ▶ The eigenvalues of  $A - \mu I$  are  $\lambda_i - \mu$ .
- ▶ The eigenvalues of  $(A - \mu I)^{-1}$  are  $\frac{1}{\lambda_i - \mu}$ .

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**Algorithm: Shifted Inverse Iteration**

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**Input:** Nonzero unit vector  $\mathbf{x}^{(0)}$

**for**  $k = 0, 1, \dots$  **until convergence do**

    Solve  $(\mathbf{A} - \mu \mathbf{I})\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)}$

    Normalise  $\mathbf{x}^{(k+1)}$

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## Rayleigh Quotient

$$\mathcal{R}_A(\mathbf{x}) = \frac{\mathbf{x}^\top \mathbf{A} \mathbf{x}}{\mathbf{x}^\top \mathbf{x}} = \arg \min_{\mu \in \mathbb{C}} \|\mathbf{A} \mathbf{x} - \mu \mathbf{x}\|$$

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Eigenvector approximation from  
eigenvalue approximation?

► Shifted Inverse Iteration

Eigenvalue approximation from  
eigenvector approximation?

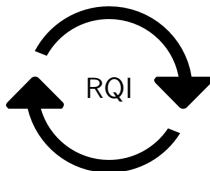
► Rayleigh Quotient

Eigenvector approximation from  
eigenvalue approximation?

► Shifted Inverse Iteration

Eigenvalue approximation from  
eigenvector approximation?

► Rayleigh Quotient



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**Algorithm:** Rayleigh Quotient Iteration

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**Input:** Nonzero unit vector  $\mathbf{x}^{(0)}$ 

$$\mu^{(0)} \leftarrow (\mathbf{x}^{(0)})^\top \mathbf{A} \mathbf{x}^{(0)}$$

**for**  $k = 0, 1, \dots$  **until convergence do**

$$\text{Solve } (\mathbf{A} - \mu^{(k)} \mathbf{I}) \mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} \text{ for } \mathbf{x}^{(k+1)}$$

Normalise  $\mathbf{x}^{(k+1)}$ 

$$\mu^{(k+1)} \leftarrow (\mathbf{x}^{(k+1)})^\top \mathbf{A} \mathbf{x}^{(k+1)}$$

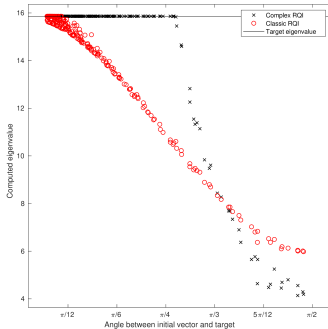
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$$\mathcal{R}_A(\mathbf{x}) = \frac{\mathbf{x}^\top \mathbf{A} \mathbf{x}}{\mathbf{x}^\top \mathbf{x}}$$

- ▶ Cubic convergence
- ▶ Converges for almost all starting vectors

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Convergence behaviour can be erratic



## Observation

Convergence depends on initial shift but **not** initial vector

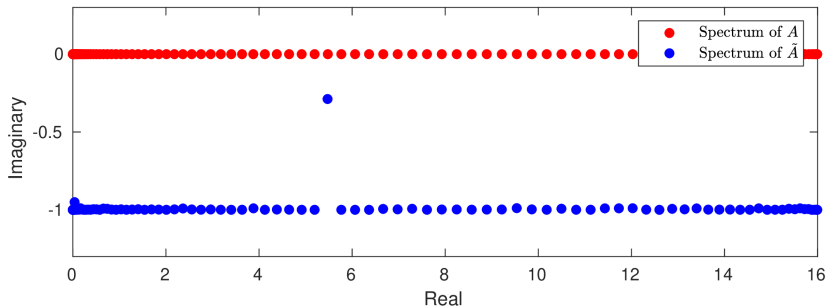


Suppose  $\mathbf{u} \approx \mathbf{v}_k$ .

$$\mathbf{A} \longrightarrow \mathbf{A} - i\gamma(\mathbf{I} - \mathbf{u}\mathbf{u}^\top) =: \tilde{\mathbf{A}}$$

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**Algorithm: Complex RQI (first version)**

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**for**  $k = 0, 1, \dots$  **until convergence do**

    Solve  $(\mathbf{A} - i\gamma(\mathbf{I} - \mathbf{u}\mathbf{u}^\top) - \mu^{(k)}\mathbf{I}) \mathbf{x}^{(k+1)} = \mathbf{x}^{(k)}$

    Normalise  $\mathbf{x}^{(k+1)}$

$\mu^{(k+1)} \leftarrow (\mathbf{x}^{(k+1)})^* \mathbf{A} \mathbf{x}^{(k+1)}$

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**Algorithm: Complex RQI (first version)**

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**for**  $k = 0, 1, \dots$  **until convergence do**

$$\text{Solve } \left( \mathbf{A} - i\gamma^{(k)} \left( \mathbf{I} - \mathbf{x}^{(k)}(\mathbf{x}^{(k)})^* \right) - \mu^{(k)} \mathbf{I} \right) \mathbf{x}^{(k+1)} = \mathbf{x}^{(k)}$$

Normalise  $\mathbf{x}^{(k+1)}$ 

$$\mu^{(k+1)} \leftarrow (\mathbf{x}^{(k+1)})^* \mathbf{A} \mathbf{x}^{(k+1)}$$

$$\gamma^{(k+1)} \leftarrow \|(\mathbf{A} - \mu^{(k+1)} \mathbf{I}) \mathbf{x}^{(k+1)}\|$$

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## Theorem

*It suffices to use*

$$\tilde{A} = A - i\gamma^{(k)}I$$

*instead of*

$$\tilde{A} = A - i\gamma^{(k)}(I - \mathbf{x}^{(k)}(\mathbf{x}^{(k)})^*).$$

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**Algorithm:** Complex Rayleigh Quotient Iteration

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**Input:** Nonzero unit vector  $\mathbf{x}^{(0)}$ 

$$\mu^{(0)} \leftarrow (\mathbf{x}^{(0)})^T \mathbf{A} \mathbf{x}^{(0)}$$

$$\gamma^{(0)} \leftarrow \|(\mathbf{A} - \mu^{(0)} \mathbf{I}) \mathbf{x}^{(0)}\|$$

$$\sigma^{(0)} \leftarrow \mu^{(0)} + i\gamma^{(0)}$$

**for**  $k = 0, 1, \dots$  **until convergence** **do**

$$\text{Solve } (\mathbf{A} - \sigma^{(k)} \mathbf{I}) \mathbf{x}^{(k+1)} = \mathbf{x}^{(k)}$$

Normalise  $\mathbf{x}^{(k+1)}$ 

$$\mu^{(k+1)} \leftarrow (\mathbf{x}^{(k+1)})^* \mathbf{A} \mathbf{x}^{(k+1)}$$

$$\gamma^{(k+1)} \leftarrow \|(\mathbf{A} - \mu^{(k+1)} \mathbf{I}) \mathbf{x}^{(k+1)}\|$$

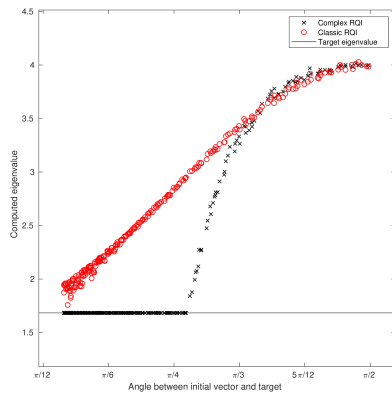
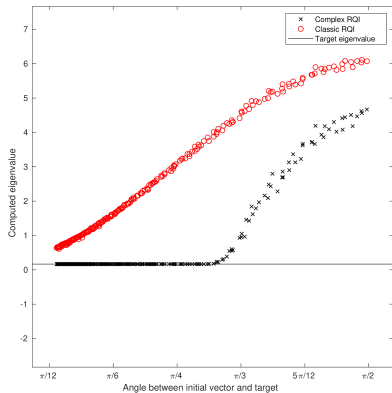
$$\sigma^{(k+1)} \leftarrow \mu^{(k+1)} + i\gamma^{(k+1)}$$

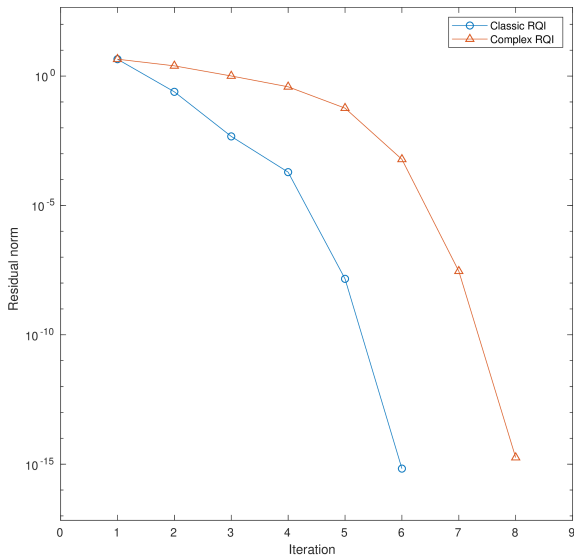
$$\mathbf{x} \leftarrow \text{Re}(\mathbf{x}^{(k+1)})$$

$$\mathbf{x} \leftarrow \mathbf{x} / \|\mathbf{x}\|$$

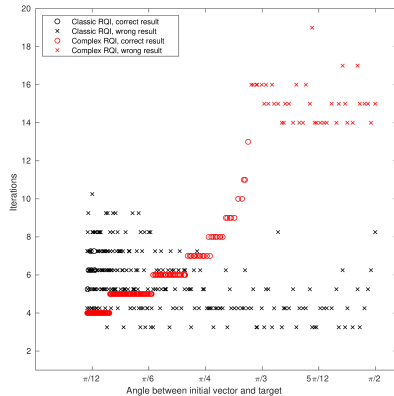
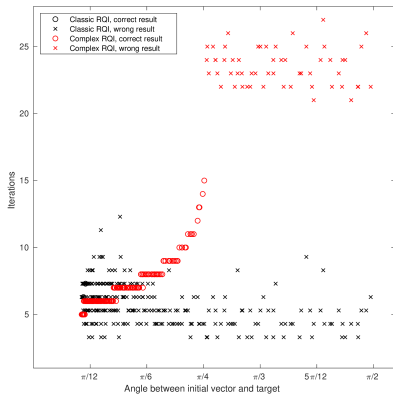
$$\mu \leftarrow \mathbf{x}^T \mathbf{A} \mathbf{x}$$

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# Summary

- ▶ Unpredictability of classic RQI
- ▶ Complex RQI succeeds for error angles below  $45^\circ$
- ▶ More iterations than classic RQI but still fast
- ▶ Theoretical analysis
- ▶ Other test problems

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Write  $\tilde{\mathbf{A}} = \mathbf{A} - i\gamma(\mathbf{I} - \mathbf{u}\mathbf{u}^\top) = \tilde{\mathbf{A}}_{(0)} + \tilde{\mathbf{A}}_{(1)}$  with

$$\tilde{\mathbf{A}}_{(0)} := \mathbf{A} - i\gamma(\mathbf{I} - \mathbf{v}_k\mathbf{v}_k^\top)$$

$$\tilde{\mathbf{A}}_{(1)} := i\gamma(\mathbf{u}\mathbf{u}^\top - \mathbf{v}_k\mathbf{v}_k^\top)$$

### Theorem

$$\lambda_j(\tilde{\mathbf{A}}) = \lambda_j(\mathbf{A}) + i\gamma \left( \langle \mathbf{v}_j, \mathbf{u} \rangle^2 - 1 \right) + \mathcal{O} \left( \left\| \tilde{\mathbf{A}}_{(1)} \right\|^2 \right),$$

where

$$\left\| \tilde{\mathbf{A}}_{(1)} \right\| = \gamma \sqrt{1 - \langle \mathbf{u}, \mathbf{v}_k \rangle^2}.$$

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### Theorem

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