

Introduction to Arnoldi method

SF2524 - Matrix Computations for Large-scale Systems

Main eigenvalue algorithms in this course

- Fundamental eigenvalue techniques (Lecture 1)
- Arnoldi method (Lecture 2-3).

- QR-method (Lecture 8-9).

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- Arnoldi method (Lecture 2-3).
Typically suitable when
 - ▶ we are interested in a small number of eigenvalues,
 - ▶ the matrix is large and sparse
 - ▶ Solvable size on current desktop $m \sim 10^6$ (depending on structure)
- QR-method (Lecture 8-9).
Typically suitable when
 - ▶ we want to compute all eigenvalues,
 - ▶ the matrix does not have any particular easy structure.
 - ▶ Solvable size on current desktop $m \sim 1000$.

Agenda lecture 2

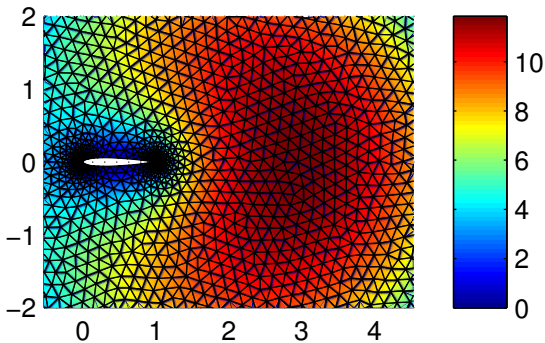
- Introduction to Arnoldi method
- Gram-Schmidt - efficiency and roundoff errors
- Derivation of Arnoldi method
- (Next lecture: Convergence characterization)

Large-scale example - finite-element method (1/2)

Laplace eigenvalue problem

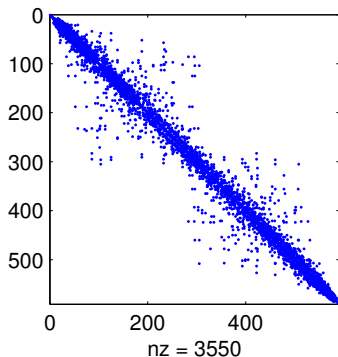
$$\Delta u = \lambda u$$

with Dirichlet boundary conditions outside wing profile.



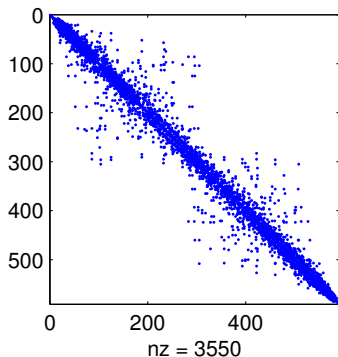
Large-scale example - finite-element method (2/2)

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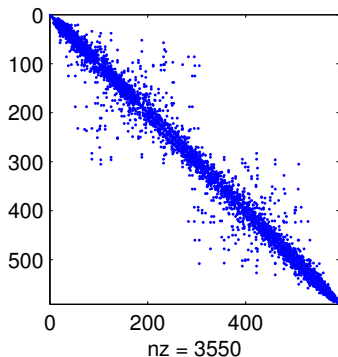


In MATLAB with very efficient QR-method:

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tic; eig(A); toc
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More than 10 hours. ⇒ need a different method

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$$Ax = \lambda x.$$

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$$Q^T A Q z = \mu z$$

are called *Ritz pairs*.

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- approximates eigenvalues of A “well” if

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Recall: Power method approximates the largest eigenvalue well.

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$$\mathcal{K}_n(A, q_1) := (q_1, Aq_1, \dots, A^{n-1}q_1)$$

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Justification of Arnoldi method

- Use Rayleigh-Ritz on $Q = (q_1, \dots, q_n)$ and $Q^T Q = I$, where

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- Arnoldi method is a “clever” procedure to construct $H_n = Q^T A Q$.
- “Clever”: We expand Q with one row in each iteration
⇒ Iterate until we are happy.

Arnoldi method graphically

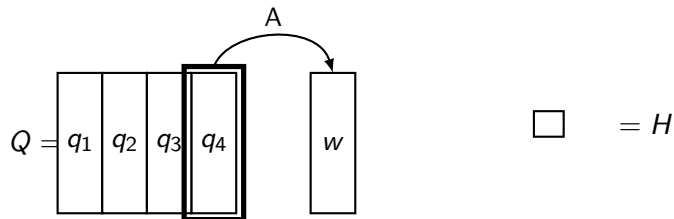
Graphical illustration of algorithm:

$$Q = \begin{array}{|c|c|c|c|} \hline q_1 & q_2 & q_3 & q_4 \\ \hline \end{array}$$

$$\square = H$$

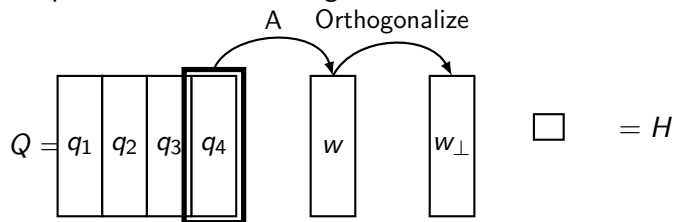
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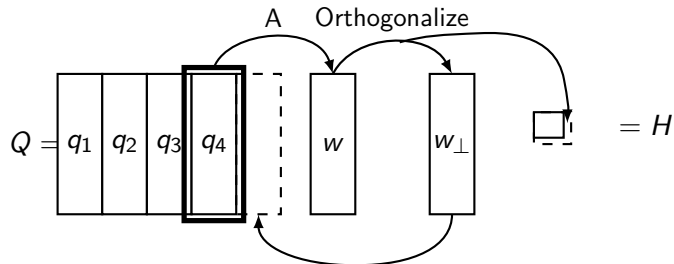
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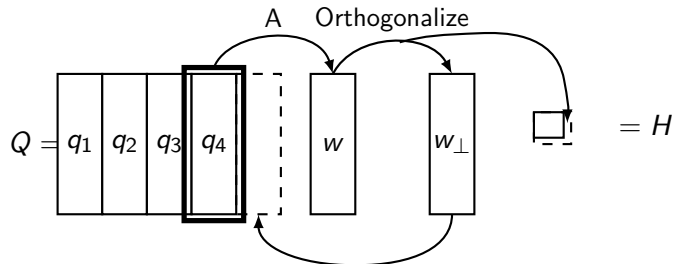
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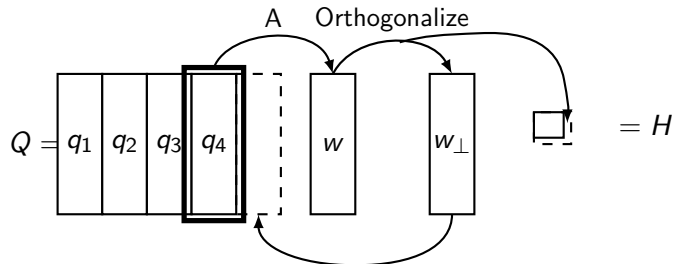
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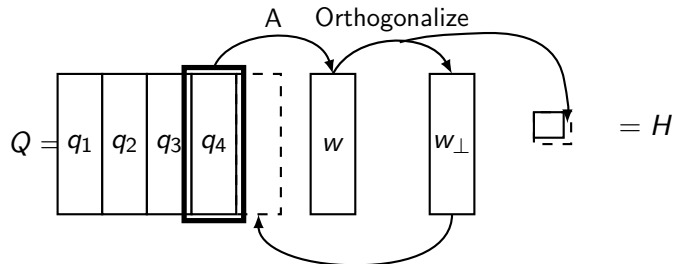
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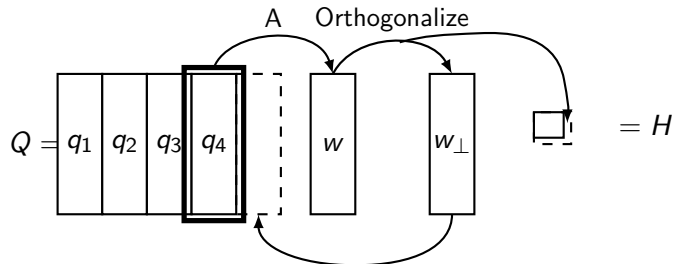
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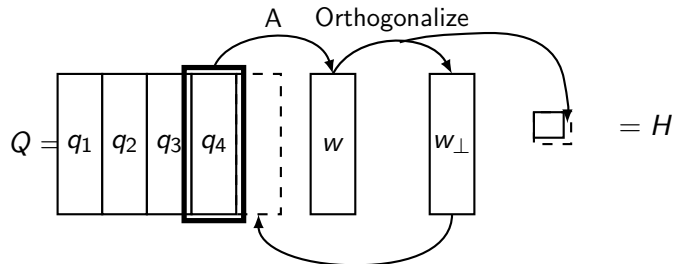


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* show arnoldi.m and Hessenberg matrix in matlab *

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We will now...

- (1) derive a good orthogonalization procedure: variants of Gram-Schmidt,
- (2) show that Arnoldi generates a Rayleigh-Ritz approximation,
- (3) characterize the convergence (next lecture?).