Introduction to Arnoldi method SF2524 - Matrix Computations for Large-scale Systems

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#### Agenda lecture 2

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

#### 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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- Fundamental eigenvalue techniques (Lecture 1)
- Arnoldi method (Lecture 2-3). Typically suitable when
  - we are interested in a small number of eigenvalues,
  - the matrix is large and sparse
  - $\sim$  Currently solvable size on 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 desktop  $m \sim 1000$ .

#### Finite-element method example (slide 1/2)

Laplace eigenvalue problem

$$\Delta u = \lambda u$$

with Dirichlet boundary conditions outside wing profile.



#### Finite-element method example (slide 2/2)

 $\Rightarrow$  Problem: Compute eigenvalues of matrix  $A \in \mathbb{R}^{m \times m}$  where *m* large but few non-zero elements:



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More than 10 hours  $\Rightarrow$  we need a different method.

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are called *Ritz pairs*.

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#### Justification of Arnoldi method

• Use Rayleigh-Ritz on  $Q = (q_1, \ldots, q_n)$  where

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Graphical illustration of algorithm:



$$\Box = H$$

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After iteration: Take eigenvalues of H as approximate eigenvalues.



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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,
- show that Arnoldi generates a Rayleigh-Ritz approximation, (2)
- (3)characterize the convergence.