

- All office hours cancelled Thu
- Exam grades ? → Wed
- Exam 1 grades 118 not 120

# QR Iteration/QR Algorithm

$$\begin{aligned}\bar{X}_0 &= A \\ \bar{Q}_0 \bar{R}_0 &= \bar{X}_0 \\ \bar{X}_1 &= \bar{R}_0 \bar{Q}_0\end{aligned}$$

$$\begin{aligned}A^k &= \bar{Q}_0 \bar{Q}, \bar{R}_1 \bar{R}_0 \\ &\vdots \\ A^k &= \dots\end{aligned}$$

→ equivalent to orth. iter

$$\bar{X}_{k+1} = \bar{Q}_k^H \bar{Q}_{k-1}^H \dots \bar{Q}_0^H A \underbrace{\bar{Q}_0 \dots \bar{Q}_k}_{Q_{all}}$$

$$Q \in \mathbb{R}^{n \times n}$$

$$Q \text{ orthogonal} \Leftrightarrow Q^T Q = I$$

$$Q \in \mathbb{C}^{n \times n}$$

$$Q \text{ unitary} \Leftrightarrow Q^H Q = I$$

## Proof sketch: Equivalence of QR iteration / Orth. iteration

### Orthogonal Iteration (no bars)

- ▶  $X_0 := A$ 
  - ▶  $Q_0 R_0 := X_0$ ,
  - ▶ where we may choose  $Q_0 = \bar{Q}_0$
  - ▶  $\hat{X}_0 = Q_0^H A Q_0 = Q_0^H Q_0 R_0 Q_0 = R_0 Q_0$
- ▶  $X_1 := A Q_0$ 
  - ▶  $Q_1 R_1 := X_1$ ,  
and because of  $X_1 = Q_0 Q_0^H A Q_0 = Q_0 \bar{X}_1 = Q_0 \bar{Q}_1 \bar{R}_1$   
we may choose  $Q_1 = Q_0 \bar{Q}_1 = \bar{Q}_0 \bar{Q}_1$ .
- ▶  $\vdots$

### QR Iteration (with bars)

- ▶  $\bar{X}_0 := A$ 
  - ▶  $\bar{Q}_0 \bar{R}_0 := A$
- ▶  $\bar{X}_1 := \bar{R}_0 \bar{Q}_0 = \hat{X}_0$ 
  - ▶  $\bar{Q}_1 \bar{R}_1 := \bar{X}_1$
- ▶  $\bar{X}_2 := \bar{R}_1 \bar{Q}_1$ 
  - ▶  $\bar{X}_2 = Q_1^H A Q_1 = \hat{X}_1$
- ▶  $\vdots$

Demo: QR Iteration [cleared]

## QR Iteration: Forward *and* Inverse

QR iteration may be viewed as performing **inverse iteration**. How?

$$\bar{X}_0 = A$$

$$X_0^{-H} = A^{-H} \quad \text{lower triangular}$$

$$\bar{Q}_k \bar{R}_k = \bar{X}_k$$

$$(\bar{Q}_k \bar{R}_k)^{-H} = \bar{Q}_k \bar{R}_k^{-H} = \bar{X}_k^{-H} \quad \text{QR factorization}$$

$$\bar{X}_{k+1} = \bar{R}_k \bar{Q}_k$$

$$\bar{X}_{k+1}^{-H} = \bar{R}_k^{-H} \bar{Q}_k$$

inverse conjugate  $((\cdot)^{-1})^H = (-)^{-H}$

## QR Iteration: Incorporating a Shift

How can we accelerate convergence of QR iteration using shifts?

$$\bar{Q}_k \bar{R}_k = \bar{X}_k - \sigma_k I \quad (\Leftrightarrow) \quad \bar{R}_k = \bar{Q}_k^H \bar{X}_k - \bar{Q}_k^H \sigma_k I$$

$$\bar{X}_{k+1} = \bar{R}_k \bar{Q}_k + \sigma_k I$$

$$\begin{aligned} \bar{X}_{k+1} &= \bar{R}_k \bar{Q}_k + \sigma_k I = (\bar{Q}_k^H \bar{X}_k - \bar{Q}_k^H \sigma_k I) \bar{Q}_k + \sigma_k I \\ &= \bar{Q}_k^H \bar{X}_k \bar{Q}_k \underbrace{\bar{Q}_k^H \bar{Q}_k}_{I} + \cancel{\sigma_k I} \\ &= \bar{Q}_k^H \bar{X}_k \bar{Q}_k \end{aligned}$$

Demo: QR Iteration [cleared] (Shifted)

## QR Iteration: Computational Expense

A full QR factorization at each iteration costs  $O(n^3)$ —can we make that cheaper?

The diagram shows the equation  $A = Q \left( \begin{array}{c|c} & \\ \hline & \end{array} \right) Q^H$ . The matrix inside the parentheses is a square with a diagonal line from the top-left to the bottom-right, representing an upper Hessenberg matrix. An arrow labeled 'H' points to this matrix. To the right, the text 'upper  $\triangle + I$ ' and 'upper Hessenberg' is written. Below the diagram, the text 'sketch QR with H' is written. Further down, it says 'QR factorization in QR it:  $O(n^2)$ ' and '...? is this shape maintained?'. On the right side, there are two equations:  $\bar{Q}_k \bar{R}_k = X_k$  and  $\bar{X}_{k+1} = \bar{R}_k \bar{Q}_k$ .

Demo: Householder Similarity Transforms [cleared]

## QR/Hessenberg: Overall procedure

Overall procedure:

1. Reduce matrix to Hessenberg form
2. Apply QR iteration using Givens QR to obtain Schur form

Why does QR iteration *stay* in Hessenberg form?

Assume  $\bar{X}_n$  is Upper Hess

$$\bar{Q}_n \bar{R}_n = \bar{X}_n \quad \bar{Q}_n = \bar{X}_n R^{-1} = \square \quad \curvearrowright \text{uH}$$
$$\bar{X}_{k+1} = \bar{R}_k \bar{Q}_k \Rightarrow \bar{X}_{n+1} \text{ also uH.}$$

What does this process look like for symmetric matrices?

## Krylov space methods: Intro

What subspaces can we use to look for eigenvectors?

QR / ortho iter:  $\text{span} (A^0 y_1, \dots, A^k y_1)$

Krylov space

$\text{span} \left( \begin{array}{c} x_0 \\ \uparrow \\ x_0 \end{array}, \begin{array}{c} A x_0 \\ \uparrow \\ x_1 \end{array}, \begin{array}{c} A^2 x_0 \\ \uparrow \\ x_2 \end{array}, \dots, \begin{array}{c} A^{k+1} x_0 \\ \uparrow \\ x_{k+1} \end{array} \right)$

$$K_k^y \left( \begin{array}{c} | \\ x_0 \\ | \end{array} \dots \begin{array}{c} | \\ x_{k+1} \\ | \end{array} \right) \leftarrow n \times k$$



# Krylov for Matrix Factorization

What matrix factorization is obtained through Krylov space methods?

$$AK_n = \begin{pmatrix} | & & | \\ x_1 & & x_n \\ | & & | \end{pmatrix} = K_n \begin{pmatrix} 0 & & & & 0 \\ \vdots & \ddots & & & \vdots \\ 0 & & & & 0 \end{pmatrix}$$

Assume invertibility

$$K_n \begin{pmatrix} 0 \\ \vdots \\ 0 \\ \alpha \\ \vdots \\ 0 \end{pmatrix} = x_i = K_n \begin{pmatrix} & & & & \\ & & & & \\ & & & & \\ & & & & \\ & & & & \end{pmatrix}$$

otherwise reconst, messy  $\hookrightarrow$  upper Hessenberg

$$\boxed{K_n^{-1}AK_n} = H$$

just another sim. transform to upper Hess!

## Conditioning in Krylov Space Methods/Arnoldi Iteration (I)

What is a problem with Krylov space methods? How can we fix it?

$$Q_n R_n = K_n \quad (\Leftrightarrow) \quad Q_n = K_n R_n^{-1}$$

$$Q_n^T A Q_n = R_n K_n^{-1} A K_n R_n^{-1}$$



$$= \square = H_n$$

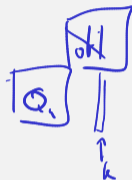
# Conditioning in Krylov Space Methods/Arnoldi Iteration (II)

$$Q = \begin{pmatrix} | \\ \bar{q}_k \\ | \end{pmatrix}$$

$$Q^T A Q = H \Leftrightarrow A Q = Q H$$

$$A \bar{q}_k = h_{1k} \bar{q}_1 + \dots + h_{k+1,k} \bar{q}_{k+1}$$

$$h_{jk} = \bar{q}_j^T A \bar{q}_k$$



Demo: Arnoldi Iteration [cleared] (Part 1)

## Krylov: What about eigenvalues?

How can we use Arnoldi/Lanczos to compute eigenvalues?



[Demo: Arnoldi Iteration \[cleared\]](#) (Part 2)