

QR iteration for eigenvalues and eigenvectors

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Recall: Last time, we linked the Gram-Schmidt process with a matrix factorization: $A = QR$, where Q is unitary (has orthonormal columns) and R is upper triangular. After explaining why the Gram-Schmidt algorithm is the poster child of instability & loss of precision, we noted that it either has to be modified (leading to the Modified and Double Gram-Schmidt algorithms), or we need to do something altogether different:

Problem: Find Q unitary (or Q a product of unitary matrices) such that $R = Q^* A$. This idea leads to two popular algorithms: Householder QR (using Householder reflectors) and Givens QR (using Givens rotations).

To get to the Householder QR algorithm, we went through the following concepts:

- **Orthogonal projector:** we say a matrix is an orthogonal projector if $P^2 = P$ and P is symmetric (hermitian). If P is a projector, $(I-P)$ is also a projector such that $R(P) = N(I-P)$ and $N(P) = R(I-P)$.
- **Orthogonal projection onto a subspace V :** given an orthonormal basis $\{q_1, \dots, q_m\}$ of V , the orthogonal projector onto V is $\sum_{i=1}^m q_i q_i^* = Q Q^*$.
- **Householder reflector:** Given a unit vector w , the Householder reflector $H_w = I - 2ww^*$. This operator "flips" the component in the direction of w , leaving its orthogonal complement y the same.

We finally derived the algorithm for Householder QR by transforming A into R one column at a time. We got $(H_{n-1} H_{n-2} \dots H_1) A = R$, which means that $A = (H_1 H_2 \dots H_{n-1}) R = QR$.

Compute IR (and maybe apply Q^*)
 • Compute IR and Q .

$$H_k = \left[\begin{array}{c|c} I_{k-1} & 0 \\ \hline 0 & I - 2\omega\omega^* \end{array} \right] \left[\begin{array}{c|c} A_{11} & A_{12} \\ \hline A_{21} & A_{22} \end{array} \right]$$

$n-(k-1)$

$$H_{k-1} A = \left[\begin{array}{c|c} A_{11} & A_{12} \\ \hline H_{k-1} A_{21} & H_{k-1} A_{22} \end{array} \right]$$

$A_{22} - 2(\underline{\underline{A_{22}\omega}})\omega$
 $O((n-(k-1))^2)$

Alg 1: (IR and apply Q^*) $\rightarrow \sim \frac{4}{3}n^3 (2GE, 1LU)$

Algo 2: (Compute \mathbf{Q}, \mathbf{R}) $\rightarrow \sim \frac{8}{3}n^3$ (4GE, 2LU)

EIGENVALUES / EIGENVECTORS

► $\mathbf{A} = \mathcal{U} \mathbf{D} \mathcal{U}^*$, \mathcal{U} unitary
 \mathbf{D} diagonal.

EF: (Eigenpair) Given $\mathbf{A} \in \mathbb{C}^{n \times n}$, then
 $\lambda \in \mathbb{C}$ is an eigenvalue of \mathbf{A} iff
there is $\vec{x} \neq 0$ s.t.

$$\mathbf{A}\vec{x} = \lambda\vec{x} \leftrightarrow \underline{(\mathbf{A} - \lambda\mathbf{I})\vec{x} = \vec{0}}$$

\vec{x} \rightarrow eigenvector (for λ)

(λ, \vec{x}) \rightarrow eigenpair of \mathbf{A}

set of λ 's \rightarrow spectrum

$(\mathbf{A} - \lambda\mathbf{I})$ $\begin{cases} \{\vec{0}\} - \lambda \text{ is not an eigen.} \\ \neq \{\vec{0}\} - \lambda \text{ is an eigenvalue.} \end{cases}$

\downarrow
 $\mathbf{A} - \lambda\mathbf{I}$ is singular - $\det(\mathbf{A} - \lambda\mathbf{I}) = 0$

$P_A(\lambda) = \det(\mathbf{A} - \lambda\mathbf{I})$ char. poly of \mathbf{A}
(deg = : n)

► eigenvalues = roots of P_A .

eigenvalues = roots of Γ_A .

otation: $E_\lambda = N(A - \lambda I)$ "eigenspace".

Multiplicity of λ : $P_A(\lambda) = (\lambda - \lambda_0)^k q(\lambda)$
 $q(\lambda_0) \neq 0.$
 $\dim(E_\lambda)$.

SIMILARITY: A, B $n \times n$, I say they are similar if $\exists P$ invertible s.t.

$$B = P^{-1} A P$$

$\triangleright A, B$ are the same linear op but P is a change of basis.

I want P to be unitary

$$B = P^* A P$$

cols of P are orthonormal.

Invariants under similarity:

- $P_A(\lambda) = P_B(\lambda)$

- Eigenvalues are the same.

(λ, \vec{x}_A) eigenpair $\leftrightarrow (\lambda, P^{-1} \vec{x}_A)$
for A eigenpair for B .

(λ, x_A) eigenpair for $A \leftrightarrow (\lambda, u^* x_A)$ eigenpair for B .

random $A \neq UDU^*$ this is not always possible.

(Schur decomposition): for $A \in \mathbb{C}^{n \times n}$, there exists a unitary U s.t.

$$T = U^* A U$$

where T is upper triangular (Schur form) and eigen of $A \rightarrow$ diagonal (T).

Algo (iterative) that produces $\approx (T, U)$

Spectral theorems (2):

Real case: $(\lambda \in \mathbb{R})$ - A be symmetric (hermitian) ($A = A^*$) \leftrightarrow there is an orthonormal basis of eigenvectors w/ $\lambda \in \mathbb{R}$, $A = UDU^*$

Complex case ($\lambda \in \mathbb{C}$) - A be normal
 $\Rightarrow AA^* = A^*A$. \leftrightarrow there is orth. basis of complex eigenvec/values $\leftrightarrow A = UDU^*$.

Sketch proof (Schur)

(i) $A = A^* \leftrightarrow T = T^* \rightarrow T = D$

$$(i) A = A^* \leftrightarrow T = T^* \rightarrow \underline{T} = \underline{D}$$

$$(ii) AA^* = A^*A \leftrightarrow T\underline{T}^* = \underline{T}^*T$$

$$\leftrightarrow \underline{T} = \underline{D}$$

Statement: There are no direct methods to find eigendecomposition / all eigenvalues of A (for $n \geq 5$)

→ If there was, you would have a formula for roots of P_A .

① QR iteration :

Start $A_0 = A$

while (algo has not converged)

$$[Q_k, R_k] = qr(A_k)$$

$$A_{k+1} = \underbrace{R_k}_{\text{ }} Q_k = \underline{Q_k^* A_k Q_k}$$

$$\dots \lim_{k \rightarrow \infty} A_k = \underline{\underline{T}}, \quad \lim_{k \rightarrow \infty} (Q_1 \dots Q_k) = \underline{\underline{Q}} =$$