

- QR iteration
- Relation to power method + simultaneous iteration

Global methods : for approximating Schur Factorization + entire spectrum.

QR algorithm:

$A_1 = Q_0 A Q_0^* \rightarrow Q_0 A Q_0^*$ is ~~Hermitian~~ reduced to upper-Hessenberg form.

~~Repeat~~

for $j=1, 2, \dots$

* Factor $A_j = Q_j R_j$

set $A_{j+1} = R_j Q_j$

end

$A_j \rightarrow$ upper Δ matrix with eigenvalues on diagonal.

Notice: $R_k = Q_k^* A_k$

$$A_{k+1} = \cancel{Q_k} R_k Q_k$$

$$= Q_k^* A_k Q_k$$

$$= \underbrace{(Q_1^* \dots Q_k^*)}_{(Q_k)^*} \cancel{A_1} \underbrace{Q_1 \dots Q_k}_{\cancel{Q_k} Q_k}$$

$$= (Q_1^* \dots Q_k^*) Q_0 A Q_0 Q_1 \dots Q_k$$

$\therefore A_{k+1}$ is unitarily similar to A .

Convergence: If $|a_1| \geq |a_2| \geq \dots \geq |a_n|$

then the p th subdiagonal entry in A_k converges to zero at rate

$$\left| \frac{a_{p+1}}{a_p} \right|^k$$

at a rate

We can understand convergence in some sense by relating the QR iteration to the power method.

~~Do Ex. 1.1~~ Suppose (for simplicity) $A \rightarrow$ upper triangular ~~matrix~~

$$A_{k+1} = (Q_k^T \dots Q_1^T) A (Q_1 \dots Q_k)$$

Next, note

$$\begin{aligned} & (Q_1 \dots Q_k)(R_1 \dots R_k) \\ &= (Q_1 \dots Q_{k-1}) Q_k R_k (R_1 \dots R_{k-1}) & \leftarrow Q_k R_k = A_k \\ &= (Q_1 \dots Q_{k-1}) A_k (R_1 \dots R_{k-1}) & \uparrow A_k = (Q_k^T \dots Q_1^T) A (Q_1 \dots Q_{k-1}) \\ &= A (Q_1 \dots Q_{k-1}) (R_1 \dots R_{k-1}) \\ &= A (Q_1 \dots Q_{k-2}) Q_{k-1} R_{k-1} (R_1 \dots R_{k-2}) & \leftarrow Q_{k-1} R_{k-1} = A_{k-1} \\ &= A (Q_1 \dots Q_{k-2}) A_{k-1} (R_1 \dots R_{k-2}) & \uparrow A_{k-1} = (Q_{k-2}^T \dots Q_1^T) A (Q_1 \dots Q_{k-2}) \\ &= A^k (Q_1 \dots Q_{k-2}) (R_1 \dots R_{k-2}) \\ &\vdots \\ &= A^k \end{aligned}$$

$\therefore (Q_1 \dots Q_k)(R_1 \dots R_k) = A^k, \quad k^{\text{th}} \text{ power of } A.$

$\tilde{Q}_k \tilde{R}_k = A^k$

Relation to symmetric / orthogonal iteration: For simplicity, suppose A is ^{Hermitian} ~~normal~~

If we start power method with $x_0 = e_1$

for $k=1, 2, \dots$
 $w = A x_{k-1}$
 $\mu_k = w / \|w\|$
 $x_k = w / \mu_k$
 etc

If we start the power method with $\hat{Q}_0 = \{e_1, e_2\}$

for $k=1, 2, \dots$

$$W_k = A \hat{Q}_{k-1}$$

$$\{\hat{Q}_k, \hat{R}_k\} = W_k \text{ \& orthonormalize}$$

end

We are running a Power iteration on multiple vectors, orthonormalizing at each step.

If $\hat{Q}_0 = I_n$:

Simultaneous iteration:

for $k=1, 2, \dots$

$$W_k = A \hat{Q}_{k-1}$$

$$\hat{Q}_k R_k = W_k$$

$$A_k = \hat{Q}_k^* A \hat{Q}_k$$

end

Thus: let \hat{q}_i^k the i th column of \hat{Q}_k : $\hat{q}_i^k \rightarrow$ vector assoc with λ_i .

Relation to QR iteration:

Simultaneous and QR iteration generate identical sequences, defined by the QR factorization of the k th power of A .

~~Power iteration~~ Shows already: for QR: $\underbrace{(Q_1 \dots Q_k)}_{\hat{Q}_k} \underbrace{(R_1 \dots R_k)}_{\hat{R}_k} = A^k$

$$\text{or } A_k = \hat{Q}_k^* A \hat{Q}_k$$

for Simultaneous iteration, show $\hat{Q}_k \hat{R}_k = A^k$, $\hat{R}_k = R_k R_{k-1} \dots R_1 \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}$

Base step ($k=0$)

$$\cancel{QA \tilde{Q} A} \quad I = A^0 = \tilde{Q}_0 \tilde{R}_0 \quad (\tilde{Q}_0 = I, \tilde{R}_0 = I \text{ for QR})$$

assume $\hat{Q}_0 = I$ for Simult.

~~In book wrong~~
Need to show for Simultaneous

If: for simultaneous, $A^{k+1} = \hat{Q}_k \hat{R}_k$

I.S.

$$A^k = A A^{k-1} = \underbrace{A \hat{Q}_{k-1}}_{W_k} \hat{R}_{k-1} = W_k \hat{R}_{k-1} = \hat{Q}_k \hat{R}_k \hat{R}_{k-1} = \hat{Q}_k \hat{R}_k$$

$$\therefore \cancel{A \tilde{Q} \tilde{R}} \quad A^k = \underbrace{\hat{Q}_k \hat{R}_k}_{\text{Simult.}} = \underbrace{\tilde{Q}_k \tilde{R}_k}_{\text{QR}}$$

$$\text{And: } A_k = \hat{Q}_k^* A \hat{Q}_k = \tilde{Q}_k^* A \tilde{Q}_k$$

\Rightarrow diagonals of A_k are the Rayleigh quotients of A com to the \tilde{q}_k vectors.

Comparing the QR iteration to the simultaneous iteration:

QR (no shift):

$$A_0 = A$$

for $k=1, 2, \dots$

$$A_{k+1} = Q_k R_k \quad (\text{factor})$$

$$A_k = R_k Q_k \quad (\text{mults})$$

$$\tilde{Q}_k = Q_1 \dots Q_k$$

Simultaneous:

$$\tilde{Q}_0 = I$$

for $k=1, 2, \dots$

$$W = A \tilde{Q}_{k-1} \quad (\text{mults})$$

$$W = \tilde{Q}_k R_k \quad (\text{in factor})$$

$$A_k = \tilde{Q}_k^* W \tilde{Q}_k$$

Both algorithms compute QR factorizations of A^k for

$$A^k = \tilde{Q}_k R_k, \quad R_k = R_1 R_2 \dots R_k$$

$$\text{and } A_k = \tilde{Q}_k^* A^k \tilde{Q}_k$$

"Simple case" A is real + symmetric, n simple eigenvalues.

$$A_k(i,i) = \tilde{Q}_k(i,i)^T A \tilde{Q}_k(i,i)$$

As $k \rightarrow \infty$, $\tilde{Q}_k(i,i) \rightarrow i^{\text{th}}$ eigenvector of A (orthog. of power method)

$A_k(i,i) = (\tilde{Q}_k(i,i))^T A (\tilde{Q}_k(i,i))$ is the Rayleigh quot that gives λ_i approx to λ_i

$$i \neq j: A_k(i,j) = \tilde{Q}_k(i,i)^T A (\tilde{Q}_k(j,j)) \rightarrow 0 \text{ by orthogonality of eigenvectors}$$

\Rightarrow If A is nonsym (non-Hermitian), and has nonsimple evs, the situation is more complicated.