

Lecture Notes 7: Eigenvalue Problem

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3 Eigenvalue Problem

3.1 Orthogonal iteration

- Want to compute more than one eigenpairs at a time.
- Algorithm(OI)
 1. $Z_0 = [\vec{v}_1, \vec{v}_2]$ is an $n \times 2$ random orthogonal matrix.
 2. For $i=1, 2, 3\dots$
 3. $Y_i = A^* Z_{i-1}$
 4. $[Z_i, W_i] = QR(Y)$
 5. end for
 6. Extract eigenpairs from Z .
 - (a) $T_k = Z_k^T A Z_k$
 - (b) if T_k is upper triangular converged ($Z_k = Q_k$)
 - (c) compute T_k 's eigenpairs.
- Suppose $X = [\vec{x}_1, \vec{x}_2, \dots, \vec{x}_n]$ eigenvector matrix of A .

Assume $|\lambda_1| > |\lambda_2| > \dots > |\lambda_n|$

$$X = QR$$

$$X^{-1} = R^{-1}Q^T$$

$$A = X\Lambda X^{-1} = QR\Lambda R^{-1}Q^T = QTQ^T \text{(Schur decomposition)}$$

$$\text{Assume } X = [x_1 x_2] = [Q_1 Q_2] \begin{bmatrix} R_{11} & R_{12} \\ 0 & R_{22} \end{bmatrix}$$

$$X^{-1} = \begin{bmatrix} R_{11}^{-1} & R_{11}^{-1}R_{12}R_{22}^{-1} \\ 0 & R_{22}^{-1} \end{bmatrix} \begin{bmatrix} Q_1^T \\ Q_2^T \end{bmatrix}$$

$$Q_1^T A Q_1 = Q_1^T X \Lambda X^{-1} Q_1$$

$$= Q_1^T [Q_1 Q_2] \begin{bmatrix} R_{11} & R_{12} \\ 0 & R_{22} \end{bmatrix} \begin{bmatrix} \Lambda_1 & \\ & \Lambda_2 \end{bmatrix} \begin{bmatrix} R_{11}^{-1} & R_{11}^{-1}R_{12}R_{22}^{-1} \\ 0 & R_{22}^{-1} \end{bmatrix} \begin{bmatrix} Q_1^T \\ Q_2^T \end{bmatrix} Q_1$$

$$= [R_{11} R_{12}] \begin{bmatrix} \Lambda_1 & 0 \\ 0 & \Lambda_2 \end{bmatrix} \begin{bmatrix} R_{11}^{-1} \\ 0 \end{bmatrix}$$

$$= R_{11} \Lambda_1 R_{11}^{-1}$$

- eigenvalues and eigenvectors of an upper triangular matrix T

$$T = \begin{bmatrix} t_{11} & t_{12} \dots & t_{1n} \\ & t_{22} \dots & t_{2n} \\ \ddots & & t_{nn} \end{bmatrix} = \begin{bmatrix} T_{11} & \vec{t}_{1i} & T_{1,i+1} \\ & t_{ii} & \vec{t}_{i,i+1}^T \\ & & T_{i+1,i+1} \end{bmatrix}$$

$$\lambda_i = t_{ii}$$

$$\vec{x}_i = \begin{bmatrix} (t_{ii}I - T_{11})^{-1} & \vec{t}_{1,n} \\ 1 & \\ 0 & \end{bmatrix}$$

- span { Z_k } \rightarrow an invariant subspace of S={ \vec{u} }

$$\vec{u} \in s \quad A\vec{u} \in s$$

$$A\vec{x}_1 = \lambda_1 \vec{x}_1$$

$$A\vec{x}_2 = \lambda_2 \vec{x}_2$$

$$\vec{u} = \alpha \vec{x}_1 + \beta \vec{x}_2$$

$$A\vec{u} = \alpha \lambda_1 \vec{x}_1 + \beta \lambda_2 \vec{x}_2$$

3.2 QR Algorithm

- Algorithm
 - To Compute all eigenpairs.
1. $B_0 = A$
 2. For $i=0,1,2\dots$
 3. $[Q_{i+1}, R_{i+1}] = qr(B_i)$
 4. $B_{i+1} = R_{i+1}Q_{i+1}, U_{i+1} = U_i Q_i$
end for
 5. Extract eigenpairs from U_K, B_K

$$\begin{aligned}
A^K Z_0 &= X \Lambda^K X^{-1} Z_0 \\
&= X \begin{bmatrix} \lambda_1^k & & & & & \\ & \lambda_2^k & & & & \\ & & \ddots & & & \\ & & & \lambda_p^k & & \\ & & & & \lambda_{p+1}^k & \\ & & & & & \ddots \\ & & & & & & \lambda_n^k \end{bmatrix} X^{-1} Z_0 \\
&= \lambda_p^k X \begin{bmatrix} (\frac{\lambda_1^k}{\lambda_p})^k & & & & & \\ & \ddots & & & & \\ & & 1 & & & \\ & & & (\frac{\lambda_{p+1}}{\lambda_p})^k & & \\ & & & & \ddots & \\ & & & & & (\frac{\lambda_n}{\lambda_p})^k \end{bmatrix} X^{-1} Z_0 \\
&\rightarrow \lambda_p^k X [X_p X_p^\perp] \begin{bmatrix} x & & & & & \\ & x & & & & \\ & & x & & & \\ & & & 1 & & \\ & & & & 0 & \\ & & & & & 0 \\ & & & & & & 0 \end{bmatrix} X^{-1} Z_0 \\
&\rightarrow \lambda_p^k (X_p W + X_p^\perp 0) \\
&[\lambda, \vec{x}_1] \quad converge \quad rate(\frac{\lambda_p}{\lambda_1})
\end{aligned}$$

— converge rate

$$\lim_{k \rightarrow \infty} = \frac{\|u_1^{(k)} - \vec{x}_1\|}{\|\vec{u}_1^{k+1} - \vec{x}_1\|} = p = \frac{1}{2} (converge \quad rate) \quad (1)$$

$$\lim_{k \rightarrow \infty} = \frac{|u_1^k - \lambda_1|}{|u_1^{k-1} - \lambda_1|} \quad (2)$$

$$\lim_{k \rightarrow \infty} = \frac{\|A\vec{u}_1^{(k)} - u_1^{(k)}\vec{u}_1^{(k)}\|}{\|A\vec{u}_1^{(k-1)} - u_1^{(k-1)}\vec{u}_1^{(k-1)}\|} \rightarrow residual \quad (3)$$