Recitation 10

Recitation Instructor: Shivam Verma

Email: shivamverma@nyu.edu

Ph: 718-362-7836

Office hours: WWH 605 (2.50 - 4.50 pm, Tuesdays)

Brief Overview

Rayleigh coefficient

Perturbation analysis

Problems from textbook

Quiz

Rayleigh coefficient

 $R(x) = \frac{x^T A x}{x^T x}$ where A is symmetric.

• If x is eigenvector, R(x) is corresponding eigenvalue

· Otherwise, if

$$oldsymbol{x} = \sum_{j=1}^n lpha_j \, oldsymbol{x}^{(j)} \, ,$$

$$R(\boldsymbol{x}) = \frac{\sum_{j=1}^{n} \lambda_j \, \alpha_j^2}{\sum_{j=1}^{n} \alpha_j^2}$$

• Th. 5.12: $\lambda_{min} \leq R(x) \leq \lambda_{max}$

• If we have a fairly close approximation of the eigenvector, R(x) gives us a good approximate of the corresponding eigenvalue!

Perturbation Analysis

The following theorems (see p. 172 of textbook) give an idea of how a matrix's eigenvalues are affected under perturbation.

Theorem 5.14 Let $M \in \mathbb{R}_{\text{sym}}^{n \times n}$, with eigenvalues λ_i and corresponding orthonormal eigenvectors $v_i, i = 1, 2, ..., n$, and suppose that $u \neq 0$ and w are vectors in \mathbb{R}^n and μ is a real number such that

$$(M - \mu I)\boldsymbol{u} = \boldsymbol{w}. \tag{5.44}$$

Then, at least one eigenvalue λ_i of M satisfies

$$|\lambda_j - \mu| \le \|{\bm w}\|_2 / \|{\bm u}\|_2$$
.

Theorem 5.15 (Bauer–Fike Theorem (symmetric case)) Suppose that $A, E \in \mathbb{R}^{n \times n}_{\text{sym}}$ and B = A - E. Assume, further, that the eigenvalues of A are denoted by $\lambda_j, j = 1, 2, ..., n$, and μ is an eigenvalue of B. Then, at least one eigenvalue λ_j of A satisfies

$$|\lambda_j - \mu| \leq ||E||_2.$$

For a general matrix, this result can be extended using the condition number of the diagonalizing/similarity transformation matrix:

 $|\lambda_i - \mu| \le k(X) ||E||_2$ where $X^{-1}AX = D$, and $k(X) = ||X^{-1}||_2 ||X||_2$ is the condition number of X.

eg.

A = [1 1/2 1/3; 1/2 1/3 1/4; 1/3 1/4 1/5]

B = [1 0.5 0.3333; 0.5 0.3333 0.25; 0.3333 0.25 0.2]

B is a perturbation (approximation) of A.

$$|\lambda_i - \mu| \le ||E||_2 = ||B - A||_2 = 3.3 \times 10^{-5}$$

In general, symmetric matrices are well-conditioned (robust to perturbation). **Exercise problems from textbook** See hints below to recall how we solved these problems in class.

- 5.3
 - Evaluate Sturm sequence at x = 0 and x = 1. Compare the sequences when $5\alpha^2 < 8$ and $5\alpha^2 > 8$.
- 5.7
 - Express R in terms of A and Q. Set this expression in $B = RQ + \mu I$ to get $B = Q^T A Q$. B is symmetric as A is symmetric.
 - Note that A is symmetric and **tridiagonal**. B is tridiagonal since $Q = \prod_{p=1 \ to \ n-1} R^{p,p+1}(\varphi)$, and each multiplication by $R^{p,p+1}(\varphi)^T$ on the left sets A(p+1,p)=0.
 - Next step is RQ, which involves taking matrix R and applying same sequence of plane rotations on the right, but with each rotation transposed. Transpose means inverse rotation (as $Q^T = Q^{-1}$), i.e. setting the element which was made 0 in previous step to non-zero.
 - Thus, B has non-zero elements below diagonal only on first sub-diagonal (<u>Hessenberg</u>).
 - o Since B is symmetric, it is tridiagonal.
 - Above steps are explained in sec. 5.7.1 in book.
- 5.8
 - $A^{(1)} = [0 \ 1; \ 1 \ 0].$ $G_{(2,1)} = [c, s; -s, c];$ we want $[a,b] = [0,1] \rightarrow [\alpha \ 0].$
 - Choose $r = \sqrt{a^2 + b^2} = 1$, c = a/r = 0, s = -b/r = -1. $G_{(2,1)} = [0, -1; 1, 0]$.
 - $\circ \quad Q = G_{(2,1)}^T \Rightarrow R = Q^T A = [-1, 0; 0, 1]$
 - $\circ \quad A^{(2)} = RQ = [-1,0; \ 0,1][0,1;-1, \ 0] = [0,1;1,0] = A^{(1)}$
 - Thus, $A^{(k)} = ... = A^{(1)}$, and A won't converge to solution. This is because our choice of $\mu = 0$. Try a different choice to see if we converge.
- 5.9
 - Follow same steps as previous question. Check your solution using MATLAB's qr function (see sample code from last recitation).
- 5.10
 - Use inverse iteration step rules to evaluate $v^{(1)}$ and $v^{(2)}$ at given value of $A, v, v^{(0)}$. Compare $v^{(2)}$ with corresponding eig-vector of A should be within 5% error. Evaluate Rayleigh coefficient at $v^{(2)}$.