

$$(A - \lambda I)x = 0$$

Ch 4: Computing eigenvalues

Problem Given A , find λ and $x \neq 0$ such that $Ax = \lambda x$

λ : e-value, frequency, growth rate, energy level
 x : e-vector, normal mode, principal component, bound state

Note We assume that A is real and symmetric. This implies that the e-values λ_i are real and the e-vectors g_i form an orthonormal basis, i.e. $g_i^T g_j = 0$ for $i \neq j$, $\|g_i\|_2 = 1$, and any x can be written as a linear combination of the g_i (Pf: omit).

$$\underline{\text{Ex}} \quad A = \begin{pmatrix} 2 & -1 \\ -1 & 2 \end{pmatrix}$$

$$f_A(\lambda) = \det(A - \lambda I) = \begin{vmatrix} 2-\lambda & -1 \\ -1 & 2-\lambda \end{vmatrix} = (2-\lambda)^2 - 1 = \lambda^2 - 4\lambda + 4 - 1 = \lambda^2 - 4\lambda + 3 = (\lambda-1)(\lambda-3)$$

$$\lambda_1 = 1 \Rightarrow (A - \lambda_1 I) x = 0 \Rightarrow \begin{pmatrix} 2-1 & -1 \\ -1 & 2-1 \end{pmatrix} x = 0 \Rightarrow \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} \sim \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} \Rightarrow \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$x_1 - x_2 = 0 \Rightarrow$ let $x_2 = s$: free parameter $\Rightarrow x_1 = x_2 = s$

$$\Rightarrow x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = s \begin{pmatrix} 1 \\ 1 \end{pmatrix}$$

$$\Rightarrow \boxed{q_1 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ 1 \end{pmatrix}}$$

$\|q_1\| = 1$

$$\lambda_2 = 3 \Rightarrow (A - \lambda_2 I) x = 0 \quad \begin{pmatrix} 2-3 & -1 \\ -1 & 2-3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$$\begin{pmatrix} -1 & -1 \\ -1 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix} \sim \begin{pmatrix} -1 & -1 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$$

$-x_1 - x_2 = 0$

Take $x_1 = 1 \Rightarrow x_2 = -1 \Rightarrow x = \begin{pmatrix} 1 \\ -1 \end{pmatrix}$

$$\|x\|_2 = \sqrt{2}$$

$$\Rightarrow \boxed{q_2 = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix}}$$

$$\|q_2\| = 1$$

Check: $q_1^T q_2 = \frac{1}{\sqrt{2}} (1 \ 1) \cdot \frac{1}{\sqrt{2}} \begin{pmatrix} 1 \\ -1 \end{pmatrix} = 0 \Rightarrow q_1 \perp q_2 \quad \checkmark$

Obvious method for computing e ' values

- Step 1 form $f_A(\lambda) = \det(A - \lambda I)$
- Step 2 solve $f_A(\lambda) = 0$ by the methods of Chapter 2 (root finding methods)

Ex $A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$, $\tilde{A} = \begin{pmatrix} 1+\epsilon & 0 \\ 0 & 1-\epsilon \end{pmatrix}$: perturbed matrix

$$f_A(\lambda) = (\lambda - 1)^2 = \lambda^2 - 2\lambda + 1 = 0 \Rightarrow \lambda = 1$$

$$f_{\tilde{A}}(\lambda) = (1 + \epsilon - \lambda)(1 - \epsilon - \lambda) = \lambda^2 - 2\lambda + 1 - \epsilon^2 = 0 \Rightarrow \lambda = 1 \pm \epsilon$$

Hence a change of size ϵ in the matrix elements leads to a change of size ϵ in the e ' values, but a change of size ϵ^2 in the coefficients of the characteristic polynomial leads to a change of size ϵ in the roots. Hence the roots of $f_A(\lambda)$ depend sensitively on the coefficients and this implies

that the obvious method for computing ϵ values is unstable.

Ex (Wilkinson)

$$A = \text{diag}(1, \epsilon, \dots, \epsilon) = \begin{pmatrix} 1 & & & \\ & \epsilon & & \\ & & \dots & \\ & & & \epsilon \end{pmatrix}$$

$$f_A = (1 - \epsilon)(\epsilon - \epsilon) \dots (\epsilon - \epsilon) = \sum_{k=0}^{\epsilon} a_k z^k \quad \epsilon_k \in (0, 1): \text{random}$$

Set $\tilde{a}_\epsilon = \epsilon (1 + 10^{-10} \epsilon_k)$,

$$p(z) = \sum_{k=0}^{\epsilon} \tilde{a}_k z^k, \quad \text{roots} = ?$$

Matlab

`plot(zeros(1, 20), 'o');` hold on

for $i = 1 : 100$

$r = \text{roots}(\text{poly}(1:20) * (\text{ones}(1, 21) + 1e-10 * \text{randn}(1, 21))));$

`plot(r, ',');` axis([0, 25, -6, 6]);

end

This example shows that roots of the characteristic polynomial can be very sensitive to perturbations in the coefficients. Hence, solving $f_A(\lambda) = 0$ is not a practical method for computing e-values (in general).

$$R_A(x) = \frac{x^T A x}{x^T x} = \frac{\langle x, Ax \rangle}{\langle x, x \rangle};$$

Def Given any $x \neq 0$, define Rayleigh Quotient

Note

$$1. \text{ For } x = g_i, \quad R_A(g_i) = \frac{g_i^T A g_i}{\underbrace{g_i^T g_i}_{=1}} = \frac{g_i^T \lambda_i g_i}{1} = \lambda_i \underbrace{g_i^T g_i}_{=1} = \lambda_i$$

2. For $x \approx g_i$, $R_A(x)$ is an approximation to λ_i and we can derive an error estimate by Taylor expansion. First recall some notation.

$$f(x_1, x_2) = f(a_1, a_2) + \frac{\partial f}{\partial x_1}(a_1, a_2)(x_1 - a_1) + \frac{\partial f}{\partial x_2}(a_1, a_2)(x_2 - a_2) + \dots$$

$\nabla f = \left(\frac{\partial f}{\partial x_1}, \dots, \frac{\partial f}{\partial x_n} \right)$: gradient

$$f(x) = f(a) + \nabla f(a) \cdot (x - a) + O(\|x - a\|^2)$$

$$R_A^*(x) = \underbrace{R_A(g_j)}_{= g_j} + \nabla R_A^*(g_j)(x - g_j) + O(\|x - g_j\|^2)$$

$$\nabla R_A^*(x) = \nabla \left(\frac{x^T A x}{x^T x} \right) = \frac{x^T x \cdot \nabla (x^T A x) - x^T A x \cdot \nabla (x^T x)}{(x^T x)^2}$$

$$x = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}$$

$$\nabla (x^T x) = \nabla (x_1^2 + x_2^2) = \begin{pmatrix} 2x_1 & 2x_2 \end{pmatrix} = 2x^T$$

$$x^T A x = (x_1 \ x_2) \begin{pmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = a_{11}x_1^2 + 2a_{12}x_1x_2 + a_{22}x_2^2$$

$$\nabla (x^T A x) = \begin{pmatrix} 2a_{11}x_1 + 2a_{12}x_2 & 2a_{12}x_1 + 2a_{22}x_2 \end{pmatrix} = 2(Ax)^T$$

$$\nabla R_A(x) = \frac{x^T x \cdot 2(Ax)^T - x^T Ax \cdot 2x^T}{(x^T x)^2} = \frac{2}{x^T x} \left((Ax)^T - R_A(x) x^T \right)$$