Chapter 4: Eigenvalue problem 固有値問題

Submit by Friday February 10th (16:00) to 数学事務・図書館 602 with the アンケート

Preliminary, 子文

Let A be an invertible matrix with n distinct eigenvalues in absolute values: $|\lambda_1| > |\lambda_2| > \cdots > |\lambda_n|$. A の絶対値固有値が相異なると想定し、それを $|\lambda_1| > |\lambda_2| > \cdots > |\lambda_n|$ と 書く。

The QR-method to approximate all the eigenvalues of the matrix A requires to compute a lot of QR-decompositions. The Gram-Schmidt orthonormalization method is slow and numerically unstable: the output matrix Q may not have orthogonal columns even for quite good approximations. A の固有値を全て近似する QR 法において、多くの QR 分解を計算する必要である。普段の Gram-Schmidt(正規) 直交化過程は遅くて数値的に不安定だと知られている:直交だと期待する行列 Q は、実際によい近似を使ってもその行が互いに直交だとは限らない。

A faster common better method is due to Francis & Kublanovskaya (around 1960): Francis 氏と Kublanovskaya 氏により、効率と数値的により安定な方法を導入した:

- Tridiagonalize the matrix A into a tridiagonal symmetric matrix T (often possible see Exercises 1, 2).
 - Aを三重対角化して、対称な三重対角行列Tを得る。TとAは同じ固有値を持つ。
- Then apply the QR -method to the matrix T instead of A. It produces only tridiagonal symmetric matrices (Exercise 3,b)) and we can apply the fast QR-decomposition for tridiagonal matrices using Givens rotations matrices (Exercise 3,a)). それから、A の代わりに三重対角行列T に QR 法を適用する。それで対称な三重対角行列の専用 QR 分解 (Givens 回転行列)を使うことができ、速くて数値的に安定あ QR 法を与える。

1 Tridiagonalization by Householder method (... 法を用いた三重対角化)

Let $u \in \mathbb{R}^n$ of norm 1 ($||u||_2 = 1$) be the Householder vector of the Householder matrix $H_u = I - 2uu^T \in \operatorname{Mat}_{n \times n}(\mathbb{R})$.

- a) Show that $H_u^T = H_u$ and $H_u^2 = I$. (Thus, H_u is orthogonal and symmetric and $H_u^{-1} = H_u$).
- b) Show that $H_u.u = -u$ and that $H_u.v = v$ if $v \in \langle u \rangle^{\perp}$ (that is H_u is the matrix of a reflection = \mathfrak{g} of axis u).

c) (Tridiagonalization=三重対角化 of a symmetric matrix. Step 1.) Write

$$A = \begin{pmatrix} a_{1,1} & a_{1,2} & a_{1,3} & \cdots \\ a_{1,2} & a_{2,2} & a_{2,3} & \cdots \\ a_{1,3} & a_{3,2} & a_{3,3} & \cdots \\ \vdots & \ddots & \ddots & \ddots \end{pmatrix}.$$

$$(1)$$

Let $s^2 = a_{1,2}^2 + \cdots + a_{1,n}^2$ with $sgn(s) = -sgn(a_{1,2})$.

Define $\vec{r} = (0, a_{1,2} - s, a_{1,3}, \dots, a_{1,n})^T \in \mathbb{R}^n$ and let $u = \vec{r} / \|\vec{r}\|_2$.

Then
$$A.H_u = \begin{pmatrix} a_{1,1} & *_{1,2} & 0 & \cdots \\ a_{1,2} & & & \\ \vdots & & B & \\ a_{1,n} & & & \end{pmatrix}$$
 with $B \in \operatorname{Mat}_{n-1,n-1}(\mathbb{R})$.

Consider the symmetric matrix $A = \begin{pmatrix} 1 & 1 & 1 & 1 \\ 1 & 2 & 2 & 2 \\ 1 & 2 & 3 & 3 \\ 1 & 2 & 3 & 4 \end{pmatrix}$.

- (i) Use the formula above to construct the vector \vec{r} and compute $||\vec{r}||_2^2$.
- (ii) Write the matrix: $\vec{r}.\vec{r}^T$.
- (iii) Compute the first line and first column *only* of the matrix product $A.(\vec{r}.\vec{r}^T)$. (don't need the whole matrix).
- (iv) Deduce that $H_u.A = A 2A\frac{\vec{r}.\vec{r}^2}{\|\vec{r}\|_2^2} = \begin{bmatrix} 1 & -\sqrt{3} & 0 & 0 \\ 1 & & B \end{bmatrix}$ (the first column is unchanged, and $B \in \text{Mat}_{3,3}(\mathbb{R})$).
- (v) Deduce that $H_u.A.H_u = \begin{bmatrix} 1 & -\sqrt{3} & 0 & 0 \\ -\sqrt{3} & & C \\ 0 & & C \end{bmatrix}$, where $C \in \operatorname{Mat}_{3,3}(\mathbb{R})$ is symmetric. (don't need to compute C).
- d) By repeating this process on the symmetric matrix $C \in \operatorname{Mat}_{3,3}(\mathbb{R})$ above, we can tridiagonalize = 三重対角化する the matrix <math>A into a tridiagonal matrix T. For a general matrix A as in (1), we need n-2 Houselholder transformations denoted H_{u_i} , $i=1,\ldots,n-2$.

$$T = H_{u_{n-2}}.H_{u_{n-3}}.\cdots.H_{u_1}.A.H_{u_1}.\cdots.H_{u_{n-3}}.H_{u_{n-2}}$$

 \rightarrow Show that H_{u_1} $H_{u_{n-2}}$ is symmetric orthogonal. Deduce that T and A have the same eigenvalues.

2 Tridiagonal matrices

Let
$$A = \begin{bmatrix} a_1 & b_1 & 0 & \cdots & 0 \\ c_1 & a_2 & b_2 & 0 & \vdots \\ & \ddots & \ddots & \ddots & \\ \vdots & 0 & c_{n-2} & a_{n-1} & b_{n-1} \\ 0 & \cdots & 0 & c_{n-1} & a_n \end{bmatrix}$$
 be a tridiagonal invertible matrix.

- a) (reduction to the symmetric case: 対称の場合に帰着する)
 - Assume that $\operatorname{sgn}(b_i) = \operatorname{sgn}(c_i)$ for all $i = 1, \ldots, n-1$ ($\operatorname{sgn} = \operatorname{sign} =$ 符号: $\operatorname{sgn}(a) = -1$ or 1 or 0 when a < 0 or a > 0 or a = 0).
 - Let $D = \begin{bmatrix} d_1 & 0 \\ 0 & d_n \end{bmatrix}$ be a diagonal invertible matrix. How to choose d_i so that $B = D.A.D^{-1}$ is tridiagonal and *symmetric*?
 - \rightarrow Why B and A have the same eigenvalues?
- b) According to Question a) we can assume A symmetric:

$$A = \begin{pmatrix} a_1 & b_1 & 0 & \cdots & 0 \\ b_1 & a_2 & b_2 & 0 & \vdots \\ 0 & \ddots & \ddots & \ddots & 0 \\ \vdots & 0 & b_{n-2} & a_{n-1} & b_{n-1} \\ 0 & \cdots & 0 & b_{n-1} & a_n \end{pmatrix}.$$
 (2)

- \rightarrow Suppose that there is a $b_i \approx 0$. How can we split the eigenvalue problem of A into two smaller ones? もしも $b_i \approx 0$ ならば、どうやって A の固有値問題をより小さな行列の二つに分解できるか。
- c) According to Questions a),b) we can assume that all b_i are non zero.
 - Let $p_k(\lambda)$ be the characteristic polynomial (固有多項式) of the k-th principal minor of A (第 k 目の主小行列). For example $p_1(\lambda) = \lambda a_1$, and $p_n(\lambda) = p_A(\lambda)$ is the characteristic polynomial of A).
 - (i) Let $p_0(\lambda)=1$. Show the recurrence relation (帰納な公式): $p_{k+1}(\lambda)=(\lambda-a_{k+1})p_k(\lambda)-b_k^2p_{k-1}(\lambda)$ holds.
 - (ii) Deduce the following three properties (we say that the sequence of polynomials $(p_0(\lambda), p_1(\lambda), \dots, p_{n-1}(\lambda), p_n(\lambda))$ is a *Sturm sequence*, or has the *Sturm property*).
 - -1- None of the polynomials $p_k(\lambda)$ is zero.
 - -2- For any $\lambda_0 \in \mathbb{C}$, for any $1 \leq k \leq n-1$, $p_k(\lambda_0) = p_{k+1}(\lambda_0)$ is impossible.
 - -3- Given $\lambda_0 \in \mathbb{R}$, if $p_k(\lambda_0) = 0$, then $\operatorname{sgn}(p_{k-1}(\lambda_0)) = -\operatorname{sgn}(p_{k+1}(\lambda_0))$.

3 QR method

Theorem 1 Given a symmetric tridiagonal invertible matrix A as in (2), with $b_i \neq 0$ for all i = 1, ..., n - 1. Its QR-decomposition A = QR verifies:

$$Q = \begin{pmatrix} q_{1,1} & q_{1,2} & \cdots & \cdots & q_{1,n} \\ q_{2,1} & q_{2,2} & q_{2,3} & \cdots & q_{2,n} \\ 0 & \ddots & \ddots & \ddots & \vdots \\ 0 & \cdots & q_{n-2,n-2} & q_{n-1,n-1} & q_{n-1,n} \\ 0 & \cdots & 0 & q_{n,n-1} & q_{n,n} \end{pmatrix} \quad R = \begin{pmatrix} z_1 & s_1 & r_1 & 0 & \cdots & 0 \\ 0 & z_2 & s_2 & r_2 & 0 & \vdots \\ 0 & \ddots & \ddots & \ddots & \ddots & 0 \\ 0 & \cdots & 0 & z_{n-2} & s_{n-2} & r_{n-2} \\ 0 & \cdots & \cdots & 0 & z_{n-1} & s_{n-1} \\ 0 & \cdots & \cdots & 0 & z_n \end{pmatrix}$$

$$(3)$$

can be computed fast using n-1 Givens rotattion matrices.

a) (the first Givens rotation P_1 on a tridiagonal matrix). Let $A = \begin{pmatrix} 3 & 1 & 0 \\ 1 & 3 & 1 \\ 0 & 1 & 3 \end{pmatrix}$

Let
$$\cos \theta_1 = \frac{a_1}{\sqrt{a_1^2 + b_1^2}} = \frac{3}{\sqrt{3^2 + 1^2}} = \frac{3}{\sqrt{10}}$$
 and $\sin \theta_1 = \frac{b_1}{\sqrt{a_1^2 + b_1^2}} = \frac{1}{\sqrt{3^2 + 1^2}} = \frac{1}{\sqrt{10}}$,

and
$$P_1 := \begin{pmatrix} \cos \theta_1 & \sin \theta_1 & 0 \\ -\sin \theta_1 & \cos \theta_1 & \\ 0 & 0 & 1 \end{pmatrix}$$
.

 \rightarrow Compute $A^{(1)}=P_1A$. We obtain a matrix $\begin{pmatrix} z_1 & s_1 & r_1 \\ 0 & x_2 & y_2 \\ 0 & 1 & 3 \end{pmatrix}$ for some values z_1, s_1, r_1, x_2, y_2 .

(参考までに 2nd Givens rotation
$$P_2$$
) By taking $\cos \theta_2 = \frac{x_2}{\sqrt{x_2^2+1^2}} \approx 0.92998$

$$\sin \theta_2 = \frac{1}{\sqrt{x_2^2 + 1^2}} \approx 0.36761, \text{ and the Givens rotation matrix } P_2 = \begin{pmatrix} 1 & 0 & 0 \\ 1 & \cos \theta_2 & \sin \theta_2 \\ 0 & -\sin \theta_2 & \cos \theta_2 \end{pmatrix},$$
 we define: $R = P_2 P_1 A = P_2 A^{(1)}$ since it is upper triangular: $R = \begin{pmatrix} \sqrt{10} & \frac{3}{2}\sqrt{10} & \frac{\sqrt{10}}{10} \\ 0 & 2.72 & 1.98 \\ 0 & 0 & 2.44 \end{pmatrix}.$

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We then have
$$Q = P_2^T.P_1^T = \begin{pmatrix} 0.95 & -0.29 & 0.12 \\ 0.32 & 0.88 & -0.35 \\ 0 & 0.37 & 0.93 \end{pmatrix}$$
.

b) In the QR-method, we define $A_0 = A$ and:

for
$$k = 0, 1, 2, ...$$
 repeat:

- 1. compute the QR-decomposition of A_k : $A_k = Q_k R_k$
- 2. define $A_{k+1} = R_k Q_k$
- \rightarrow Prove that $A_1 = R_0 Q_0$ and $A_0 = A$ have the same eigenvalues.
- c) To compute fast the QR-decomposition of A_1, A_2, A_3, \ldots by using Givens rotations, all computed matrices A_1, A_2, \ldots should be tridiagonal symmetric (as A_0)...
 - \rightarrow Prove that if Q_0 and R_0 both have the shape given in (3) of Theorem 1, then R_0Q_0 is tridiagonal symmetric.