

Problem for 2010 July

Proposed by Dan Jurca

Suppose $1 \leq k \leq n \Rightarrow x_k \in \mathbf{R}$,

$$1 \leq i \leq n, 1 \leq j \leq n \Rightarrow a_{ij} = \begin{cases} x_i x_j & \text{if } i \neq j \\ x_i x_j + 1 = x_i^2 + 1 & \text{if } i = j, \end{cases}$$

and consider the $n \times n$ matrix $A = (a_{ij})$.

- a. Find the determinant of A .
- b. Find the characteristic equation and the spectrum of A .
- c. Find n linearly independent eigenvectors of A .
- d. Find the inverse of A .

Solution by the proposer

If each $x_k = 0$, then $A = I_n$, the $n \times n$ identity matrix, so the problem is trivial. Therefore, suppose $x_m \neq 0$ for some m , $1 \leq m \leq n$. If $b_{ij} = x_i x_j$ and $B = (b_{ij}) = A - I_n$ then the spectrum of A , $\sigma(A)$, equals $\{1 + \lambda \mid \lambda \in \sigma(B)\}$. With $s = x_1^2 + x_2^2 + \dots + x_n^2$ we will show that the characteristic equation of B is $q(\lambda) = (\lambda - s)\lambda^{n-1}$. Consider the vectors $\mathbf{v}^1, \mathbf{v}^2, \dots, \mathbf{v}^n$ as follows.

$$\mathbf{v}^m = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ \vdots \\ x_n \end{bmatrix} \quad k \neq m \Rightarrow \mathbf{v}^k = \begin{bmatrix} v_1^k \\ v_2^k \\ v_3^k \\ \vdots \\ v_n^k \end{bmatrix} \quad \text{where } v_i^k = \begin{cases} -x_k & \text{if } i = m \\ x_m & \text{if } i = k \\ 0 & \text{if } i \neq m \text{ and } i \neq k \end{cases}$$

We compute $B\mathbf{v}^m$ and each $B\mathbf{v}^k$, $k \neq m$.

$$(B\mathbf{v}^m)_i = \sum_{j=1}^n b_{ij} x_j = \sum_{j=1}^n (x_i x_j) x_j = \sum_{j=1}^n x_i (x_j x_j) = x_i \sum_{j=1}^n x_j^2 = s x_i \quad \text{whence}$$

$$B\mathbf{v}^m = s\mathbf{v}^m; \quad \text{and}$$

$$k \neq m \Rightarrow (B\mathbf{v}^k)_i = \sum_{j=1}^n b_{ij} v_j^k = b_{im} v_m^k + b_{ik} v_k^k = (x_i x_m) \cdot -x_k + (x_i x_k) \cdot x_m = 0 \quad \text{whence}$$

$$B\mathbf{v}^k = \mathbf{0} = 0\mathbf{v}^k.$$

It follows that $\sigma(B) = \{s; 0\}$, and associated eigenvectors are $\mathbf{v}^m; \mathbf{v}^1, \dots, \mathbf{v}^{m-1}, \mathbf{v}^{m+1}, \dots, \mathbf{v}^n$, which are linearly independent. (It is obvious that $\{\mathbf{v}^k \mid k \neq m\}$ is a linearly independent set. Since $k \neq m \Rightarrow A\mathbf{v}^k = 0\mathbf{v}^k$ and $A\mathbf{v}^m = s\mathbf{v}^m \neq 0\mathbf{v}^m$, it follows that \mathbf{v}^m is not a linear combination of the other \mathbf{v}^k .) Hence $\sigma(A) = \{s+1; 1\}$ and associated eigenvectors are $\mathbf{v}^m; \mathbf{v}^1, \dots, \mathbf{v}^{m-1}, \mathbf{v}^{m+1}, \dots, \mathbf{v}^n$; $\det(A) = s+1$; and the characteristic polynomial of A is $p(\lambda) = (\lambda - (s+1))(\lambda - 1)^{n-1}$. Since $\det(A) = s+1 \neq 0$, A is invertible.

Next let

$$c_{ij} = \begin{cases} s+1 - x_i^2 & \text{if } i = j \\ -x_i x_j & \text{if } i \neq j, \end{cases}$$

and let $C = (c_{ij})$; we show that $AC = \det(A) \cdot I_n$.

First we compute

$$\begin{aligned}
(AC)_{ii} &= \sum_{k=1}^n a_{ik}c_{ki} \\
&= \sum_{k=1}^{i-1} a_{ik}c_{ki} + a_{ii}c_{ii} + \sum_{k=i+1}^n a_{ik}c_{ki} \\
&= \sum_{k=1}^{i-1} (x_i x_k) \cdot (-x_k x_i) + (x_i^2 + 1) \cdot (s + 1 - x_i^2) + \sum_{k=i+1}^n (x_i x_k) \cdot (-x_k x_i) \\
&= -x_i^2 \sum_{k=1}^{i-1} x_k^2 + s x_i^2 + x_i^2 - x_i^4 + s + 1 - x_i^2 - x_i^2 \sum_{k=i+1}^n x_k^2 \\
&= -x_i^2 \sum_{k=1}^n x_k^2 + s x_i^2 + s + 1 \\
&= -x_i^2 s + s x_i^2 + s + 1 \\
&= s + 1; \quad \text{and then}
\end{aligned}$$

$$\begin{aligned}
i < j \Rightarrow (AC)_{ij} &= \sum_{k=1}^n a_{ik}c_{kj} \\
&= \sum_{k=1}^{i-1} a_{ik}c_{kj} + a_{ii}c_{ij} + \sum_{k=i+1}^{j-1} a_{ik}c_{kj} + a_{ij}c_{jj} + \sum_{k=j+1}^n a_{ik}c_{kj} \\
&= \sum_{k=1}^{i-1} (x_i x_k) \cdot (-x_k x_j) + (x_i^2 + 1) \cdot (-x_i x_j) + \sum_{k=i+1}^{j-1} (x_i x_k) \cdot (-x_k x_j) + \\
&\quad + (x_i x_j) \cdot (s + 1 - x_j^2) + \sum_{k=j+1}^n (x_i x_k) \cdot (-x_k x_j) \\
&= -x_i x_j \sum_{k=1}^{i-1} x_k^2 - x_i x_j \sum_{k=i+1}^{j-1} x_k^2 - x_i x_j \sum_{k=j+1}^n x_k^2 - x_i^3 x_j - x_i x_j + s x_i x_j + x_i x_j - x_i x_j^3 \\
&= -x_i x_j [s - x_i^2 - x_j^2] - x_i^3 x_j + s x_i x_j - x_i x_j^3 \\
&= -x_i x_j [s - x_i^2 - x_j^2 + x_i^2 - s + x_j^2] \\
&= 0;
\end{aligned}$$

and similarly $j < i \Rightarrow (AC)_{ij} = 0$. Therefore $AC = (s + 1) \cdot I_n = \det A \cdot I_n$, so that, finally,

$$\begin{aligned}
A^{-1} &= \frac{1}{s + 1} C, \quad \text{where} \\
s &= x_1^2 + \cdots + x_n^2.
\end{aligned}$$

Remark. If we let $\mathbf{x} = \mathbf{v}^m = (x_1, x_2, \dots, x_n)^T$, then we have $A = \mathbf{x}\mathbf{x}^T + I_n$ and $C = (s + 2)I_n - A$. Hence, for example, $A\mathbf{x} = (\mathbf{x}\mathbf{x}^T + I_n)\mathbf{x} = \mathbf{x}(\mathbf{x}^T\mathbf{x}) + \mathbf{x} = \mathbf{x}s + \mathbf{x} = (s + 1)\mathbf{x}$, so that $s + 1 \in \sigma(A)$ with associated eigenvector \mathbf{x} . Next $AC = A((s + 2)I_n - A) = (s + 2)A - A^2 = sA + I_n - I_n + 2A - A^2 = sA + I_n - (A - I_n)^2 = sA + I_n - \mathbf{x}\mathbf{x}^T\mathbf{x}\mathbf{x}^T = sA + I_n - \mathbf{x}s\mathbf{x}^T = sA + I_n - s\mathbf{x}\mathbf{x}^T = sA + I_n - s(A - I_n) = (s + 1)I_n$, so that $A^{-1} = 1/(s + 1) \cdot C = 1/(s + 1) \cdot ((s + 2)I_n - A)$, as before.

Also solved by Matthew Felix and Massoud Malek. Massoud Malek has generalized this result.