

Engineering Mathematics Basics

Matrix Basics

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1. Let A be the $n \times n$ matrix with $a_{j,j+1} = 1, j = 1, \dots, n-1$ and all other elements zero. Represent A as a sum of vector outer products.

solution

$$A = \sum_{j=1}^{j=n-1} I_j I_j$$

2. Consider the block matrix $\begin{bmatrix} A & B \\ C & D \end{bmatrix}$ with $A \in C^{n \times n}$ nonsingular.

1) Determine $X \in C^{m \times m}$ and $Y \in C^{n \times n}$ such that

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} I & O \\ CA^{-1} & I \end{bmatrix} \begin{bmatrix} A & B \\ O & X \end{bmatrix} \text{ (blockLU factorization)}$$

and that

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} I & BD^{-1} \\ O & I \end{bmatrix} \begin{bmatrix} Y & O \\ C & D \end{bmatrix} \text{ (blockLU factorization)}$$

solution:

prove the first equation: if $X = D - CA^{-1}B$,

$$\begin{bmatrix} I & O \\ CA^{-1} & I \end{bmatrix} \begin{bmatrix} A & B \\ O & X \end{bmatrix} = \begin{bmatrix} IA & IB \\ CA^{-1}A & CA^{-1}B + IX \end{bmatrix} = \begin{bmatrix} A & B \\ C & CA^{-1}B + X \end{bmatrix} = \begin{bmatrix} A & B \\ C & D \end{bmatrix}$$

prove the second equation: if $Y = A - BD^{-1}C$,

$$\begin{bmatrix} I & BD^{-1} \\ O & I \end{bmatrix} \begin{bmatrix} Y & O \\ C & D \end{bmatrix} = \begin{bmatrix} IY + BD^{-1}C & BD^{-1}D \\ IC & ID \end{bmatrix} = \begin{bmatrix} Y + BD^{-1}C & B \\ C & D \end{bmatrix} = \begin{bmatrix} A & B \\ C & D \end{bmatrix}$$

2) Prove that

$$\det(A)\det(D - CA^{-1}B) = \det(D)\det(A - BD^{-1}C) \quad (1)$$

$$\det(I - CB) = \det(I - BC) \quad (2)$$

$$\det(I - xy^*) = 1 - y^*x \quad (3)$$

where $x, y \in C^n$.

solution:

prove the first equation:

$$\begin{aligned} \begin{bmatrix} I & O \\ CA^{-1} & I \end{bmatrix} \begin{bmatrix} A & B \\ O & X \end{bmatrix} &= \begin{bmatrix} A & B \\ C & D \end{bmatrix} = \begin{bmatrix} I & BD^{-1} \\ O & I \end{bmatrix} \begin{bmatrix} Y & O \\ C & D \end{bmatrix} \\ \det\left(\begin{bmatrix} I & O \\ CA^{-1} & I \end{bmatrix}\right)\det\left(\begin{bmatrix} A & B \\ O & X \end{bmatrix}\right) &= \det\left(\begin{bmatrix} I & BD^{-1} \\ O & I \end{bmatrix}\right)\det\left(\begin{bmatrix} Y & O \\ C & D \end{bmatrix}\right) \\ \det(I)\det(AX) &= \det(I)\det(YD) \\ \det(AX) &= \det(YD) \end{aligned}$$

$$,X = D - CA^{-1}B, Y = A - BD^{-1}C$$

$$\det(A)\det(D - CA^{-1}B) = \det(D)\det(A - BD^{-1}C)$$

prove the second equation:

if $A=D=I$,

$$\det(I)\det(I - CI^{-1}B) = \det(I)\det(I - BI^{-1}C)$$

$$\det(I - CB) = \det(I)\det(I - BC)$$

prove the last equation:

if $C=x, B=y^*$,

$$\det(I - xy^*) = \det(I)\det(I - y^*x)$$

$$\det(I - xy^*) = \det(I - y^*x)$$

Because A is a 1×1 matrix,

$$\det(I - xy^*) = 1 - y^*x$$

3. Let $A \in C^{n \times n}$, the trace of A is defined as the sum of its diagonal elements, i.e., $\text{trace}(A) = \sum_{i=1}^n a_{ii}$.

1) Show that the trace is a linear function, i.e., if $A, B \in C^{n \times n}$ and $\alpha, \beta \in C$, then

$$\text{trace}(\alpha A + \beta B) = \alpha \text{trace}(A) + \beta \text{trace}(B).$$

solution:

$$\text{trace}(\alpha A + \beta B) = \sum_{i=1}^n (\alpha a_{ii} + \beta b_{ii}) = \alpha \sum_{i=1}^n a_{ii} + \beta \sum_{i=1}^n b_{ii} = \alpha \text{trace}(A) + \beta \text{trace}(B)$$

2) Show that $\text{trace}(AB) = \text{trace}(BA)$, even though in general $AB \neq BA$.

solution:

$$\text{trace}(AB) = \sum_{i=1}^n \sum_{k=1}^n a_{ik}b_{ki}$$

$$\text{trace}(BA) = \sum_{i=1}^n \sum_{k=1}^n b_{ik}a_{ki} = \sum_{k=1}^n \sum_{i=1}^n b_{k0i0}a_{i0k0} = \sum_{k=1}^n \sum_{i=1}^n a_{i0k0}b_{k0i0} = \text{trace}(AB)$$

3) Show that if $S \in R^{n \times n}$ is skew-symmetric, i.e., $S^T = -S$, the $\text{trace}(S) = 0$. Prove the converse to this statement or provide a counterexample.

solution:

$$\text{trace}(S) = \text{trace}(S^T) = \text{trace}(-S) = -\text{trace}(S)$$

So, $\text{trace}(S)=0$.

counterexample of the converse to this statement:

$$S = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}, \text{trace}\left(\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}\right) = 0$$

$$S^T = \left(\begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix}\right)^T = \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \neq \begin{bmatrix} 0 & -1 \\ 0 & 0 \end{bmatrix} = -S$$

4. Let $\lambda_1, \lambda_2, \dots, \lambda_n$ be the eigenvalues of A . Show that

$$\det(A) = \lambda_1 \lambda_2 \dots \lambda_n, \quad (1)$$

$$\text{trace}(A) = \lambda_1 + \lambda_2 + \dots + \lambda_n, \quad (2)$$

$$\text{trace}(A^k) = \lambda_1^k + \lambda_2^k + \dots + \lambda_n^k, k = 1, 2, \dots \quad (3)$$

solution:

prove the first equation: $A = C^{-1} \begin{bmatrix} \lambda_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \lambda_n \end{bmatrix} C$

$$\det(A) = \det(C^{-1}) \det \left(\begin{bmatrix} \lambda_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \lambda_n \end{bmatrix} \right) \det(C)$$

$$\det(A) = \det(C^{-1}C) \det \left(\begin{bmatrix} \lambda_1 & \cdots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \cdots & \lambda_n \end{bmatrix} \right)$$

$$\det(A) = \lambda_1 \cdots \lambda_n$$

prove the second equation:

$$\det(\lambda - A) = (\lambda - \lambda_1) \cdots (\lambda - \lambda_n)$$

the coefficient of the term of λ^{n-1} is $-(\lambda_1 + \cdots + \lambda_n)$

$$\det(\lambda - A) = \det \left(\begin{bmatrix} \lambda - a_{1,1} & \cdots & -a_{1,n} \\ \vdots & \ddots & \vdots \\ -a_{n,1} & \cdots & \lambda - a_{n,n} \end{bmatrix} \right)$$

the coefficient of the term of λ^{n-1} is only related to $(\lambda - a_{11})(\lambda - a_{22}) \cdots (\lambda - a_{nn})$

so, the coefficient of the term of λ^{n-1} is $-(a_1 + \cdots + a_n)$

$$\text{so, } -(\lambda_1 + \cdots + \lambda_n) = -(a_1 + \cdots + a_n)$$

prove the last equation:

$$\text{if } A\alpha = \lambda_1\alpha, A^k = \lambda^k\alpha$$

So, $\lambda_1^k, \lambda_2^k, \dots, \lambda_n^k$ be the eigenvalues of A^k .

$$\text{So, } \text{trace}(A^k) = \lambda_1^k + \lambda_2^k + \dots + \lambda_n^k, k = 1, 2, \dots$$

5. 1) Let x_1, x_2, \dots, x_k be eigenvectors of A. Show that $S = \text{span}\{x_1, x_2, \dots, x_k\}$ is invariant under A.

solution:

if $y \in S$,

$$\exists a_1, \dots, a_k, y = a_1x_1 + \cdots + a_kx_k$$

$$\text{so, } Ay = a_1\lambda_1x_1 + \cdots + a_k\lambda_kx_k$$

so, S is invariant under A.

- 2) Let $A = \begin{bmatrix} 2 & 1 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$. Show that the space $S = \text{span}\{e_1, e_2\}$ is invariant under A and is not spanned

$$\text{by eigenvectors of A. Here } e_1 = \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} \text{ and } e_2 = \begin{bmatrix} 0 \\ 1 \\ 0 \end{bmatrix}.$$

solution:

$$\forall y \in S, y = c_1 e_1 + c_2 e_2$$

$$Ay = Ac_1 e_1 + Ac_2 e_2 = c_1 A e_1 + c_2 A e_2 = c_1 \begin{bmatrix} 2 \\ 0 \\ 0 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ 2 \\ 0 \end{bmatrix} = (2c_1 + c_2)e_1 + 2c_2 e_2$$

so, S is invariant under A .

$$\det(\lambda E - A) = 0 \text{ can derive } \lambda_1 = \lambda_2 = 2, \lambda_3 = 1$$

$$(2E - A)\alpha = 0 \text{ can derive } \alpha_1 = k_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix}$$

$$(E - A)\alpha = 0 \text{ can derive } \alpha_2 = k_2 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

$$e_2 = b_1 \begin{bmatrix} 1 \\ 0 \\ 0 \end{bmatrix} + b_2 \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix} \text{ can derive that } b_1, b_2 \text{ has no solution.}$$

so, S is not spanned by eigenvectors of A .

6. Verify that $\|xy^*\|_F = \|xy^*\|_2 = \|x\|_2 \|y\|_2$ for any $x, y \in C^n$.

solution:

$$\|xy^*\|_F^2 = \text{trace}((yx^*)(xy^*)) = (x^*x)\text{trace}(yy^*) = \sum_{i=1}^n x_i^2 \sum_{j=1}^n y_j^2$$

$$\|xy^*\|_2 = \max_z ((\|xy^*z\|_2^2) / (\|z\|_2^2)) = \max_z ((|y^*z|^2 \|x\|_2^2) / (\|z\|_2^2)) = \|y\|_2^2 \|x\|_2^2 / \|y\|_2^2 = \|x\|_2^2$$

$\|x\|_2^2$, k is a number.

7. Let $A \in C^{n \times m}$. Show that $\|A\|_2 \leq \|A\|_F$. (Recall that $\text{trace}(B) = \sum_i \lambda_i(B)$).

solution:

$$\|A\|_F^2 = \text{trace}(A^*A)$$

$$\|A\|_2^2 = \max_x \frac{\|Ax\|_2^2}{\|x\|_2^2} = \max_i \lambda_i$$

$$\text{trace}(A^*A) = \sum_{i=1}^n \lambda_i > \max_i \lambda_i$$

8. For $A \in C^{n \times m}$, show that

$$\|A\|_1 = \max_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1} = \max_{1 \leq j \leq n} \sum_{i=1}^m |a_{ij}| \quad (1)$$

$$\|A\|_\infty = \max_{x \neq 0} \frac{\|Ax\|_\infty}{\|x\|_\infty} = \max_{1 \leq j \leq n} \sum_{i=1}^m |a_{ij}| \quad (2)$$

solution:

$$\|A\|_1 = \max_{x \neq 0} \frac{\|Ax\|_1}{\|x\|_1} = \max_{x \neq 0} \frac{\sum_i |\sum_j a_{ij} x_j|}{\sum_i |x_i|} \leq \max_{x \neq 0} \frac{\sum_i \sum_j |a_{ij} x_j|}{\sum_i |x_i|} \leq \max_{x \neq 0} \frac{(\max_j \sum_i |a_{ij}|) \sum_j |x_j|}{\sum_i |x_i|} = \max_j \sum_i |a_{ij}|$$

The proof of the second equation is similar.

9. Let $A, B \in R^{m \times n}$. Recall that $\|A\|_F^2 = \text{trace}(A^T A)$ and that for any matrix D for which the product AD is defined, $\text{trace}(AD) = \text{trace}(DA)$.

1) Show that the Q that minimizes $\|A - QB\|_F$ over all choices of orthogonal Q also maximizes $\text{trace}(A^T QB)$.

solution:

$$\|A - QB\|_F^2 = \text{trace}((A - BQ)^T(A - QB)) = \text{trace}((A^T - B^T Q^T)(A - QB)) = \text{trace}(A^T A + B^T Q^T QB - A^T QB - B^T Q^T A) = \text{trace}(A^T A + B^T B) - 2\text{trace}(A^T QB)$$

$\text{trace}(A^T A + B^T B)$ is constant.

so, the Q that minimizes $\|A - QB\|_F$ over all choices of orthogonal Q also maximizes $\text{trace}(A^T QB)$

2) Suppose that the SVD of the $m \times m$ matrix BA^T is $U\Sigma V^T$, where U and V are $m \times m$ and orthogonal and Σ is diagonal with diagonal entries $\sigma_1 \geq \dots \geq \sigma_m \geq 0$. Define $Z = V^T QU$. Use these definitions and (i) to show that

$$\text{trace}(A^T QB) = \text{trace}(Z\Sigma) \leq \sum_{i=1}^m \sigma_i$$

.

solution:

$$\text{trace}(A^T QB) = \text{trace}(QBA^T) = \text{trace}(QU\Sigma V^T) = \text{trace}(V^T QU\Sigma) = \text{trace}(Z\Sigma)$$

Z is orthogonal and diagonal, hence $z_{ij} = 0$ or 1 or -1 , $\text{trace}(Z\Sigma) = \sum_i z_{ii}\sigma_i \leq \sum_i \sigma_i$

3) Identify the choice of Q that gives equality in the bound of (ii).

solution:

if $Q = VU^T$,

$$Z = V^T QU = V^T VU^T U = V^{-1} VU^{-1} U = II = I$$

$$\text{trace}(Z\Sigma) = \text{trace}(\Sigma) = \sum_i \sigma_i$$

4) Carefully state a theorem summarizing the solution to the minimization of $\|A - QB\|_F$.

solution:

$Q = VU^T$, $Q^T = Q^{-1}$, $\|A - QB\|_F$ is minimized by $Q \Rightarrow BA^T = U\Sigma V^T$ is an SVD.

10. Let $A \in \mathbb{C}^{m \times n}$.

1) Show that $(\text{range}(A))^\perp = \text{null}(A^*)$ and that $(\text{null}(A))^\perp = \text{range}(A^*)$.

solution:

$$A = U\Sigma V^*, r = \text{Rank}(A) \Rightarrow \text{null}(A^*) = \text{span}\{u_{r+1}, \dots, u_m\}, \text{range}(A) = \text{span}\{u_1, \dots, u_r\} \Rightarrow (\text{range}(A))^\perp = \text{null}(A) \Rightarrow \text{range}(A) = (\text{null}(A))^\perp$$

2) Show respectively that AA^+ , A^+A , $I - A^+A$ and $I - AA^+$ are the orthogonal projectors onto $\text{range}(A)$, $\text{range}(A^*)$, $\text{null}(A)$ and $\text{null}(A^*)$.

solution:

First part:

$$(AA^+)^2 = (AA^+A)A^+ = AA^+$$

$$\forall x \in \text{range}(A), x = Ay = AA^+Ay = AA^+x \Rightarrow x \in \text{range}(AA^+)$$

$$\forall x \in \text{range}(AA^+), x = AA^+y = A(A^+y) \Rightarrow x \in \text{range}(A)$$

hence AA^+ are the orthogonal projectors onto $\text{range}(A)$.

Second part:

By first part, $A^*(A^*)^+$ are the orthogonal projectors onto $\text{range}(A^*)$.

hence A^+A are the orthogonal projectors onto $\text{range}(A^*)$.

Third part:

By first part, $I - AA^+$ are the orthogonal projectors onto $\text{range}(A^\perp) = \text{null}(A)$.

hence $I - AA^+$ are the orthogonal projectors onto $\text{null}(A)$.

Last part:

By second part, $I - A^+A$ are the orthogonal projectors onto $\text{range}((A^*)^\perp) = \text{null}(A^*)$.

hence $I - A^+A$ are the orthogonal projectors onto $\text{null}(A^*)$.

3) Suppose $\text{rank}(A) = r$, let $A = U\Sigma V^*$ with $U = [U_1 \ U_2]$ and $V = [V_1 \ V_2]$ unitary and partitioned, such that $U_1 \in C^{m \times r}$ and $V_1 \in C^{n \times r}$. Show that $U_1U_1^*, V_1V_1^*, V_2V_2^*$ and $U_2U_2^*$ are the orthogonal projectors onto $\text{range}(A), \text{range}(A^*), \text{null}(A)$ and $\text{null}(A^*)$, respectively.

solution:

$$AA^+ = U\Sigma V^*(V\Sigma^+U^*) = U \begin{bmatrix} I_r & 0 \\ 0 & 0 \end{bmatrix} U^* = U_1U_1^*$$

hence $U_1U_1^*$ are the orthogonal projectors onto $\text{range}(A)$.

other parts are similar.