

VV256 Honors Calculus IV

Fall 2022 — Problem Set 6



December 5, 2022

Note: All of the solutions in this problem set is written by me on notability, but my bad handwriting may cause a lot of typo. Anyway, thanks TA Huang for turning my awful handwriting into this Latex file which is more easily for you to read.

Exercise 6.1

Given $A' \in \text{Mat}_{(n-1) \times n}(\mathbb{C})$, $B' \in \text{Mat}_{n \times (n-1)}(\mathbb{C})$, show that

$$\det(A'B') = \sum_{j=1}^n \det(M_j) \det(N_j)$$

where $M_j \in \text{Mat}_{n-1}(\mathbb{C})$ is A' with j -th column deleted, and $N_j \in \text{Mat}_{n-1}(\mathbb{C})$ is B' with j -th row deleted.

Answer:

Note: this is the application of cauchy-Binet Theorem. Detailed proof follows:
Let $v \in \mathbb{C}^n$ given by

$$v := \begin{bmatrix} (-1)^{n+1}M_1 & \cdots & (-1)^{n+j}M_j & \cdots & (-1)^{2n}M_n \end{bmatrix}^\top$$

then $v \in \ker A'$, since for $k = 1, \dots, n$, by Laplace expansion,

$$(A'v)_k = \sum_j a_{kj}(-1)^{n+j}M_j = \det \left(\begin{bmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & & \vdots \\ a_{k1} & \cdots & a_{kn} \\ \vdots & & \vdots \\ a_{n-1,1} & \cdots & a_{n-1,n} \\ \hline a_{k1} & \cdots & a_{kn} \end{bmatrix} \right) = 0$$

thus $A'v = 0$. If $v = 0$ (all largest minors are zero), then $\text{rank } A \leq n - 2$, thus

$$\text{rank}(AB) \leq \min\{\text{rank}(A), \text{rank}(B)\} \leq n - 2$$

Note that $AB \in M_{n-1}(\mathbb{C})$, thus $\text{rank}(AB) \leq n - 2$ implies $\det(AB) = 0$. On the other hand, note that $v = 0$ implies that $\det(M_j) = 0$ for $j = 1, \dots, n$. Therefore LHS = RHS = 0 in (B.66). If $v \neq 0$, let

$$A := \begin{bmatrix} A' \\ -v^\top - \end{bmatrix} \in M_n(\mathbb{C}), \quad B := \begin{bmatrix} B' & \begin{vmatrix} | \\ | \\ | \end{vmatrix} \\ v & \begin{vmatrix} | \\ | \\ | \end{vmatrix} \end{bmatrix} \in M_n(\mathbb{C})$$

hence M_j is the (n, j) -minor of A , and N_j the (j, n) -minor of B , and

$$\det(AB) = \det \left(\left[\begin{array}{c|c} A'B' & A'v \\ \hline v^*B' & v^*v \end{array} \right] \right) = \det \left(\left[\begin{array}{c|c} A'B' & 0 \\ \hline v^*B' & v^*v \end{array} \right] \right) = \|v\|^2 \det(A'B')$$

Similarly by Laplace expansion of $\det(A)$,

$$\det(A) = \sum_{j=1}^n |M_j|^2 = \|v\|^2$$

and of $\det(B)$ (note that \pm signs cancel)

$$\det(B) = \sum_{j=1}^n M_j N_j$$

The rest of the proof follows from $\det(AB) = \det(A) \det(B)$.

Exercise 6.2

Given a matrix $A \in \text{Mat}_n(\mathbb{C})$, the adjugate of A denoted by $\text{adj } A \in M_n(\mathbb{C})$ is given by $(\text{adj } A)_{ij} := (-1)^{i+j} \det(a_{mk})_{m \neq j, k \neq i}$. Show that

- (i) Let $A \in \text{Mat}_n(\mathbb{C})$, then $A(\text{adj } A) = (\text{adj } A)A = (\det A)I_n$.
- (ii) Let $A, B \in \text{Mat}_n(\mathbb{C})$, then $\text{adj}(AB) = \text{adj}(B) \text{adj}(A)$.
- (iii) If $X \in \text{Mat}_n(\mathbb{C})$ is invertible, then $\text{adj}(X^{-1}YX) = X^{-1} \text{adj}(Y)X$, for all $Y \in \text{Mat}_n(\mathbb{C})$.
- (iv) Given $A \in \text{Mat}_n(\mathbb{C})$ with $\text{rank}(A) \leq n-2$, then $\text{adj } A \equiv 0 \in \text{Mat}_{n-1}(\mathbb{C})$.
- (v) Given $A \in \text{Mat}_n(\mathbb{C})$ with $\text{rank}(A) = n-1$, then $\text{adj } A = \alpha vw^\top$ for some scalar α , where $v \in \ker A$, and $w \in \ker A^\top$.

Answer:

- (i) $A \cdot (\text{adj } A)$

$$\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ \vdots & \vdots & & \vdots \\ a_{n1} & a_{n2} & \dots & a_{nn} \end{pmatrix} \begin{pmatrix} A_{11} & A_{21} & \dots & A_{n1} \\ A_{12} & A_{22} & \dots & A_{n2} \\ \vdots & & & \vdots \\ A_{1n} & A_{2n} & \dots & A_{nn} \end{pmatrix}$$

Note that the product of an element in A and the cofactor of other element in other lines will always be zero.

$$\begin{aligned}
&= \begin{pmatrix} \sum_{i=1}^n a_{1i} A_{1i} & 0 & \cdots & 0 \\ 0 & \sum_{i=1}^n a_{2i} A_{2i} & \cdots & c_0 \\ 0 & \cdots & \cdots & \sum_{i=1}^n a_{ni} A_{ni} \end{pmatrix} \begin{matrix} A_{ij} \text{ is the algebraic cofactor} \\ \text{of } a_{ij} \end{matrix} \\
&= \begin{pmatrix} |A| & 0 & 0 \\ 0 & |A| & \\ 0 & \cdots & |A| \end{pmatrix} = |A| \cdot I_n \text{ proved}
\end{aligned}$$

You can verify $\text{adj}(A) \cdot A = A \cdot \text{adj}(A)$ Yourself using this property.

(ii) $\text{adj}(AB) = \text{adj}(B) \cdot \text{adj}(A)$ if A, B both invertible. (**Only do this is wrong!**)

$$\text{adj}(B) = B^{-1} \det(B)$$

$$\text{adj}(A) = A^{-1} \det(A)$$

$$\therefore \text{adj}(B) \cdot \text{adj}(A) = (B^{-1} A^{-1}) \det(AB) = (AB)^{-1} \det(AB) = \text{adj}(AB)$$

if they are both not invertible. (It means at least one of A or B is not invertible)

You can refer to the brief proof in my rc slide because the following proof may have typo.

$$\text{adj}(A)_{ij} = (-1)^{i+j} \det(a_{mk})_{m \neq j, k \neq i}$$

$$\text{adj}(A) = \begin{pmatrix} A_{11} & A_{21} & \cdots & A_{n1} \\ A_{12} & A_{22} & \cdots & A_{n2} \\ \vdots & \vdots & & \\ A_{1n} & A_{2n} & \cdots & A_{nn} \end{pmatrix}$$

$$\text{adj}(A)_{ij} = A_{ji} = (-1)^{i+j} a_{ij} M_{ij}$$

$$(AB)_{ij} = \sum_k a_{ik} b_{kj}$$

$$AB = \begin{pmatrix} \sum_k a_{1k} b_{k1} & \cdots & \sum_k a_{nk} b_{k1} \\ \sum_k a_{1k} b_{k2} & \cdots & \sum_k a_{nk} b_{k2} \\ \vdots & & \vdots \\ \sum_k a_{1k} b_{kn} & \cdots & \sum_k a_{nk} b_{kn} \end{pmatrix}$$

$$\text{adj}(AB)_{ij} = \text{Algebra cofactor } ((AB)_{ij})$$

$$= (-1)^{i+j} \det \left(\sum_l a_{ml} b_{lk} \right) m \neq i, k \neq j$$

Note that $(\sum_l a_{ml} b_{lk}) m \neq i, k \neq j$ is the matrix of $n-1$ order, refer to your 6.1, You will find

$$\det \left(\sum_l a_{ml} b_{lk} \right) m \neq i, k \neq j = \sum_l \det(M_l) \cdot \det(N_l)$$

$$\begin{aligned} M_l &= (a_{ms})_{m \neq j, s \neq l}, N_l = (b_{nk})_{k \neq i, n \neq l} \\ &= \sum_l \det(a_{ms})_{m \neq j, s \neq l} \cdot \det(b_{nk})_{k \neq i, n \neq l} (-1)^{i+j} \\ &= \sum_l (-1)^{j+l} \det(a_{ms})_{n \neq j, s \neq l} \cdot (-1)^{i+l} \det(b_{nk})_{k \neq i, n \neq l} \\ &= \sum_l A_{lj} \cdot B_{jl} \\ &= \sum_l B_{jl} \cdot A_{lj} \\ &= \sum_i (\text{adj } B)(\text{adj } A) \text{ . Proved.} \end{aligned}$$

(iii)

$$\begin{aligned} &\text{adj}(X^{-1}YX) \\ &= \text{adj}(YX) \text{adj}(X^{-1}) \\ &= \text{adj}(X) \text{adj}(Y) \text{adj}(X^{-1}) \\ &= X^{-1} \det X \text{adj}(Y) X \det(X^{-1}) \\ &= X^{-1} \text{adj}(Y) X \quad \text{proved} \end{aligned}$$

(iv)

$$A = \begin{pmatrix} a_{11} & \dots & a_{1n} \\ a_{21} & \dots & a_{2n} \\ \vdots & & \vdots \\ a_{n1} & \dots & \dots \end{pmatrix}$$

with rank $n - 2$, we can perform

$$\text{adj } A = \begin{pmatrix} A_{11} & \dots & A_{n1} \\ A_{12} & & \vdots \\ \vdots & & \vdots \\ A_{1n} & \dots & A_{nn} \end{pmatrix}$$

the first minors will be 0 thus $\text{adj } A \equiv 0$. Note, the rank of n matrix determine the order of the highest order non-zero minor, when $r \leq n - 2$, the highest order non-zero minor will be at most $n - 2$, thus the adj matrix consist of $n - 1$ order minus will be the zero matrix

(v) $A(\text{adj } A) = |A| \cdot I_n$ by (i) since $\text{rank}(A) = n - 1$, $|A| = 0$. ($\text{rank}(A) = n \Leftrightarrow |A| \neq 0$)

$$\therefore A(\text{adj } A) = 0$$

$$\therefore \text{adj } A \in \ker A.$$

By rank-nullity theorem.

$$\dim \ker A + \text{rank}(A) = n$$

$$\dim \ker A = 1$$

$$\text{So, } \ker A = \text{range}(\text{adj } A)$$

$$\therefore v \in \ker A$$

$$\therefore \text{range}(\text{adj } A) = \text{span}(v)$$

$$A^\top (\text{adj } A)^\top = ((\text{adj } A)A)^\top = 0$$

$$\therefore (\text{adj}(A))^\top \in \ker A^\top$$

Also by rank-nullity theorem

$$\ker A^\top = \text{range}(\text{adj } A^\top)$$

$$\therefore w \in \ker A^\top$$

$$\therefore \text{range}(\text{adj}(A)^\top) = \text{span}(w)$$

Since $\text{rank}(A) = n - 1$, So, $\text{adj } A \neq 0$.

so. $\text{rank}(\text{adj}(A)) \geq 1$

$$\frac{\text{rank}(A)}{n-1} + \frac{\text{rank}(\text{adj}(n))}{?} \leq n - \text{rank}(A \cdot \text{adj}(A))$$

$$\text{rank}(\text{adj}(A)) \leq 1.$$

$$\therefore \text{rank}(\text{adj}(A)) = 1$$

\therefore It can be expressed as $\alpha v w^T$, no α also ok.

Exercise 6.3

Given an invertible matrix function $A(t) \in \text{Mat}_n(\mathbb{R})$ with differentiable entries, verify that

$$\frac{d}{dt} [A(t)^{-1}] = -A(t)^{-1} \frac{dA(t)}{dt} A(t)^{-1}$$

(i) by using the identity $\mathbf{D}\iota(A) \cdot H = -A^{-1}HA^{-1}$, where $\iota(A) = A^{-1}$.

(ii) by differentiating $A(t)A(t)^{-1} = I_n$.

Answer:

(i) set the direction matrix to the direction of t we can set it triuclly.

(ii)

$$\frac{d}{dt} [A(t)A(t)^{-1}] = \frac{d}{dt} (I_n)$$

$$\left(\frac{d}{dt} A(t) \right) \cdot A(t)^{-1} + \frac{d}{dt} \cdot A(t)^{-1} \cdot A(t) = 0$$

$$\frac{d}{dt} A^{-1}(t) = -A(t)^{-1} \cdot \frac{d}{dt} A(t) A(t)^{-1}$$

proved.

Exercise 6.4

Let $\Phi(t) \in \text{Mat}_n(\mathbb{R})$ satisfy the IVP

$$\frac{d\Phi(t)}{dt} = A(t)\Phi(t), \quad \Phi(0) = I_n$$

where $A(t) \in \text{Mat}_n(\mathbb{R})$ for all $t \in I \subseteq \mathbb{R}$. Assume all calculations are possible,

- (i) Verify $\Phi(-t) = \Phi(t)^{-1}$ by showing both sides satisfy the same IVP.
- (ii) Show that

$$\det \Phi(t) = \exp \left(\int_0^t \text{tr } A(s) ds \right)$$

Answer:

(i)

This problem is wrong. Refer to the problem in hmwk7

(ii) directly proved by Liouville formula. or by example 4.11

$$\frac{d}{dt} \det \phi(t) = \sum_{i=1}^n \begin{vmatrix} y_{11} & y_{12} & \cdots & y_{1n} \\ \vdots & & & \\ y'_{ii} & \cdots & & y'_{in} \\ \vdots & & & \\ y_{n1} & \cdots & y_{nn} \end{vmatrix}$$

$$\phi(t) = \begin{vmatrix} y_{11}(t) & \cdots & y_{1n}(t) \\ \vdots & & \vdots \\ y_{n1}(t) & \cdots & y_{nn}(t) \end{vmatrix}$$

$$= \sum_{i=1}^n \begin{vmatrix} y_{11} & \cdots & y_{1n} \\ \sum_{j=1}^n a_{ij} y_{j1} & \cdots & \cdots \\ y_{n1} & \cdots & y_{nn} \end{vmatrix}$$

$$= \sum_{i=1}^n a_{ii} \det \phi(t)$$

$$= \text{tr}(A(t)) \det \phi(t)$$

$$\frac{1}{\det \phi(t)} d(\det \phi(t)) = \text{tr}(A(t)) dt$$

$$\ln \det \phi(t) = \int \text{tr}(A(s)) ds$$

$$\begin{aligned} \det \phi(t) &= \det(\phi(t_0)) e^{\int_{t_0}^t \text{tr } A(s) ds} \\ &= \det(\phi(t_0)) e^{\int_0^t \text{tr } A(s) ds} \text{ proved} \end{aligned}$$

You can also refer to my RC in which I use Der 3 to prove this by chain rule.

Exercise 6.5

Given a smooth vector field $f : \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}^n$, suppose $u : \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}^n$, $(x, t) \mapsto u(x, t)$ satisfies

$$\frac{\partial}{\partial t} u(x, t) = f(u(x, t), t), \quad u(x, 0) = x$$

Let the Jacobian determinant be given by

$$J(x, t) = \det \left(\frac{\partial u}{\partial x} \right) = \det \left(\frac{\partial u_i}{\partial x_j} \right) = \det \left(\begin{bmatrix} \frac{\partial u_1}{\partial x_1} & \cdots & \frac{\partial u_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial u_n}{\partial x_1} & \cdots & \frac{\partial u_n}{\partial x_n} \end{bmatrix} \right)$$

and the divergence of f in the variable u is given by

$$\operatorname{div} f(u, t) = \sum_{k=1}^n \frac{\partial f_k}{\partial u_k}(u, t)$$

show that

$$\frac{\partial}{\partial t} J(x, t) = \operatorname{div} f(u, t) \cdot J(x, t)$$

Answer:

$$\begin{aligned} \frac{\partial}{\partial t} J(x, t) &= \frac{\partial}{\partial t} \det \begin{pmatrix} \frac{\partial u_1}{\partial x_1} & \cdots & \frac{\partial u_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial u_n}{\partial x_1} & \cdots & \frac{\partial u_n}{\partial x_n} \end{pmatrix} \\ &= \sum_{i=1}^n \det \begin{pmatrix} \frac{\partial u_1}{\partial x_1} & \cdots & \frac{\partial u_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial u_i}{\partial x_1 \partial t} & \cdots & \frac{\partial u_i}{\partial x_n \partial t} \\ \vdots & & \vdots \\ \frac{\partial u_n}{\partial x_1} & \cdots & \frac{\partial u_n}{\partial x_n} \end{pmatrix} \\ &= \sum_{i=1}^n \det \begin{pmatrix} \frac{\partial u_1}{\partial x_1} & \cdots & \frac{\partial u_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial f(u_i, t)}{\partial x_1} & \cdots & \frac{\partial f(u_i, t)}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial u_n}{\partial x_1} & \cdots & \frac{\partial u_n}{\partial x_n} \end{pmatrix} \\ &= \sum_{i=1}^n \det \begin{pmatrix} \frac{\partial u_1}{\partial x_1} & \cdots & \frac{\partial u_1}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial f(u_i, t)}{\partial x_1} & \cdots & \frac{\partial f(u_i, t)}{\partial x_n} \\ \vdots & & \vdots \\ \frac{\partial u_n}{\partial x_1} & \cdots & \frac{\partial u_n}{\partial x_n} \end{pmatrix} \\ &= \sum_{i=1}^n \frac{\partial f(u_i, t)}{\partial u_i} \det(\cdot) \\ &= \sum_{i=1}^n \frac{\partial f(u_i, t)}{\partial u_i} J(x, t) \text{ proved} \end{aligned}$$

Exercise 6.6

Let $A \in \text{Mat}_n(\mathbb{R})$, let $f_k(A) = A^k, k \in \mathbb{N}$. Show that

$$\mathbf{D} \operatorname{tr}(f_k(A)) \cdot H = k \operatorname{tr}(A^{k-1}H)$$

Answer:

$$\begin{aligned} \operatorname{tr}(Df_k(A)H) &= \operatorname{tr}\left(\left.\frac{d}{dt}\right|_{t=0} (A + tH)^k\right) \\ &= \operatorname{tr}\left(\sum_{m=0}^{k-1} A^m H A^{k-1-m}\right) \\ &= \operatorname{tr}\left(\sum_{m=0}^{k-1} A^{k-1} H\right) \\ &= k \operatorname{tr}(A^{k-1}H) \\ &= \left.\frac{d}{dt}\right|_{t=0} \operatorname{tr}(f_k(A + tH)) \\ &= \left.\frac{d}{dt}\right|_{t=0} \operatorname{tr}((A + tH)^k) \\ &= \left.\frac{d}{dt}\right|_{t=0} \sum_{m=0}^{k-1} \operatorname{tr}(A^m H A^{k-1-m}) \\ &= \left.\frac{d}{dt}\right|_{t=0} \operatorname{tr}\left(\sum_{m=0}^{k-1} A^m H A^{k-1-m}\right) \\ &= k \operatorname{tr}(A^{k-1}H) \end{aligned}$$

Exercise 6.7

Recall that

$$\exp(A) := \sum_{n \geq 0} \frac{A^n}{n!}$$

for a square matrix A .

(i) show that

$$\mathbf{D} \exp(A) \cdot H = \sum_{n \geq 0} \frac{1}{(n+1)!} \sum_{k=0}^n A^k H A^{n-k}$$

(ii) In particular, show that

$$\operatorname{tr}(\mathbf{D} \exp(A) \cdot H) = \operatorname{tr}(\exp(A)H)$$

(iii) Show that

$$\mathbf{D} \exp(A) \cdot H = \int_0^1 e^{(1-t)A} H e^{tA} dt$$

Answer:

(i)

$$\begin{aligned} \text{Dexp}(A) \cdot H &= \left. \frac{d}{dt} \right|_{t=0} \exp(A + tH) \\ &= \left. \frac{d}{dt} \right|_{t=0} I + \left. \frac{d}{dt} \frac{(A + tH)}{1} \right|_{t=0} + \left. \frac{d}{dt} \frac{(A + tH)^2}{2!} \right|_{t=0} + \dots + \left. \frac{d}{dt} \frac{(A + tH)^{n+1}}{(n+1)!} \right|_{t=0} \end{aligned}$$

You perform derivative on $t = 0$, the only thing remain will be $t \cdot \sum_{k=0}^n A^k H A^{n-k}$ (something like binomial expansion) derivative it will remain $\sum_{k=0}^n A^k H A^{n-k}$, coefficient is easy to derive thus $\text{Dexp}(A)H = \sum_{n=0}^{\infty} \frac{1}{(n+1)!} \sum_{k=0}^n A^k H A^{n-k}$ proved

(ii)

$$\begin{aligned} \exp(A)H &= \sum_{n=0}^{\infty} \frac{A^n}{n!} H \\ &= H + AH + \frac{1}{2}A^2H + \frac{1}{6}A^3H + \dots \\ \text{Dexp}(A)H &= H + \frac{1}{2}[HA + AH] + \frac{1}{6}[A^2H + AHA + HA^2] + \dots \end{aligned}$$

You shall count how many terms and you will find you have proved this.

According to the property

$$\text{tr}(A) + \text{tr}(B) = \text{tr}(A + B)$$

$$\text{tr}(AB) = \text{tr}(BA)$$

$$\begin{aligned} \text{tr Dexp}(A) &= \text{tr}(H) + \frac{1}{2}[\text{tr}(HA) + \text{tr}(AH)] \\ &\quad + \frac{1}{6}[\text{tr}(A^2H) + \text{tr}(AHA) + \text{tr}(HA^2)] + \dots \\ &= \text{tr}(H) + \text{tr}(AH) + \frac{1}{2}\text{tr}(A^2H) + \dots \\ &= \text{tr exp}(A) \quad \text{proved.} \end{aligned}$$

$$\begin{aligned}
& \int_0^1 e^{(1-t)A} H e^{tA} dt \\
&= \int_0^1 \sum_{n=0}^{\infty} \frac{[(1-t)A]^n}{n!} H \sum_{m=0}^{\infty} \frac{(tA)^m}{m!} \\
&= \int_0^1 \sum_{n=0}^{\infty} \frac{(1-t)^n A^n}{n!} H \cdot \sum_{m=0}^{\infty} \frac{t^m A^m}{m!} dt \\
&= \sum_{m \geq 0} \sum_{n \geq 0} \frac{1}{m!n!} \int_0^1 t^m (1-t)^n dt A^m H A^n \\
&= \sum_{m \geq 0} \sum_{n \geq 0} \frac{1}{m!n!} \frac{\Gamma(m+1)\Gamma(n+1)}{\Gamma(m+n+2)} A^m H A^n \\
&= \sum_{m \geq 0} \sum_{n \geq 0} \frac{1}{m!n!} \frac{m!n!}{(m+n+1)!} A^m H A^n \\
&= \sum_{m \geq 0} \sum_{n \geq 0} \frac{1}{(m+n+1)!} A^m H A^n \\
&= \sum_{m \geq 0} \frac{1}{(m+1)!} \sum_{n \geq 0}^m A^n H A^{m-n} \text{ proved}
\end{aligned}$$

Reference

1. Runze Cai, Notes on The Eigenvector-Eigenvalue Identity Paper (2021)
2. QiHong Xie, Advanced Algebra Fudan University
3. Tongren Ding Ordinary Differential Equation Tutorial Beijing University