

Project Option One

Liu Mingdao 20200111156

Problem 2.1

Note that

$$\begin{aligned}\frac{d}{dt}(e^{-At}y) &= -e^{-At}Ay + e^{-At}\frac{dy}{dt} \\ &= e^{-At}\left(\frac{dy}{dt} - Ay\right) \\ &= e^{-At}f(t, y)\end{aligned}$$

Integral from 0 to t on both sides, we have

$$\begin{aligned}\int_0^t d(e^{-As}y) &= \int_0^t e^{-As}f(s, y)ds \\ e^{-At}y(t) - e^{-A \cdot 0}y(0) &= \int_0^t e^{-As}f(s, y)ds \\ y(t) - e^{At}c &= \int_0^t e^{A(t-s)}f(s, y)ds\end{aligned}$$

Finally

$$y(t) = e^{At}c + \int_0^t e^{A(t-s)}f(s, y)ds$$

□

Problem 2.2

We know that both $f(x) = \cos(xt)$ and $g(x) = x^{-1}\sin(xt)$ are even functions of x . Note that f and g are analytic functions, so $f(x)$ and $g(x)$ can be expanded at $x = 0$ as power series with only even powers of x . Then $f(\sqrt{A})$ and $g(\sqrt{A})$ are just the power series of A , which is totally independent from the choice of \sqrt{A} .

Therefore the value of $y(t) = f(\sqrt{A})y_0 + g(\sqrt{A})y'_0$ doesn't depend on the choice of \sqrt{A} , so it can be unique. □

Problem 2.3

First we decompose A into Jordan normal form

$$A = Z \begin{bmatrix} J_1 & \\ & J_2 \end{bmatrix} Z^{-1}$$

where the eigenvalues of J_1 is in the left half plane and the eigenvalues of J_2 in the right half plane. Partition Z as $[Z_1 \ Z_2]$. We have

$$W = \frac{1}{2}(\text{sign}(A) + I) = [O \ Z_2]Z^{-1}$$

Let V_+ be the sum of all A -invariant subspaces whose eigenvalues have a positive real part, V_- be the sum of all A -invariant subspaces whose eigenvalues have a negative real part. From Jordan canonical form, we know

Z_1 form a basis of V_- , Z_2 form a basis of V_+ . Since $W = [O \ Z_2]Z^{-1}$, the linearly independent columns of W form a basis of V_+ , i.e. $\text{Ran}(W) = V_+$. Partition Q as $[Q_1 \ Q_2]$. Then we get

$$W = Q_1[R_{11} \ R_{12}]\Pi^T$$

So the columns of Q_1 form a basis of $\text{Ran}(W)$, which is V_+ . Then $AQ_1 = Q_1B$ for some $B \in M_q(\mathbb{C})$.

Then we know by direct calculation

$$Q^T A Q = \begin{bmatrix} Q_1^T A Q_1 & Q_1^T A Q_2 \\ Q_2^T A Q_1 & Q_2^T A Q_2 \end{bmatrix} = \begin{bmatrix} B & Q_1^T A Q_2 \\ Q_2^T Q_1 B & Q_2^T A Q_2 \end{bmatrix} = \begin{bmatrix} B & Q_1^T A Q_2 \\ O & Q_2^T A Q_2 \end{bmatrix} := \begin{bmatrix} A_{11} & A_{12} \\ & A_{22} \end{bmatrix}$$

For any eigenpair (λ, x) of B (which is A_{11}), we have $Bx = \lambda x$, thus $AQ_1x = Q_1Bx = \lambda Q_1x$, so (λ, Q_1x) is an eigenpair of A . Note that $Q_1x \in V_+$, we have $\text{Re}(\lambda) > 0$.

Note that A have no eigenvalue on the imaginary axis, so $\mathbb{C}^n = V_- \oplus V_+$. Thus, the columns of Q_2 form a basis of V_- . Let $AQ_2 = Q_2C$ for some $C \in M_{(n-q)}(\mathbb{C})$. Note that $C = Q_2^T A Q_2 = A_{22}$. For any eigenpair (μ, y) of C , we have $Cy = \mu y$, thus $AQ_2y = Q_2Cy = \mu Q_2y$, so (μ, Q_2y) is an eigenpair of A . Note that $Q_2y \in V_-$, we have $\text{Re}(\mu) < 0$. \square

Problem 2.4

First show that $A\#B = A^{\frac{1}{2}}B^{\frac{1}{2}}$. Notice that

$$\begin{aligned} (A^{\frac{1}{2}}B^{-\frac{1}{2}})^2 &= A^{\frac{1}{2}}B^{-\frac{1}{2}}A^{\frac{1}{2}}B^{-\frac{1}{2}} \\ &= A^{\frac{1}{2}}A^{\frac{1}{2}}B^{-\frac{1}{2}}B^{-\frac{1}{2}} \\ &= AB^{-1} \end{aligned}$$

$A^{\frac{1}{2}}$ and $B^{-\frac{1}{2}}$ commute, so they can be simultaneously diagonalized. Let $A^{\frac{1}{2}} = X\Lambda_A X^{-1}$, $B^{-\frac{1}{2}} = X\Lambda_B X^{-1}$, we know $A^{\frac{1}{2}}B^{-\frac{1}{2}} = X\Lambda_A\Lambda_B X^{-1}$, $A^{\frac{1}{2}}B^{-\frac{1}{2}}$ is positive definite. Note that the principal square root of a matrix is unique. So $A^{\frac{1}{2}}B^{-\frac{1}{2}} = (AB^{-1})^{\frac{1}{2}}$. Thus, $A\#B = (AB^{-1})^{\frac{1}{2}}B = A^{\frac{1}{2}}B^{-\frac{1}{2}}B = A^{\frac{1}{2}}B^{\frac{1}{2}}$.

Then show that for this hermitian positive definite A , $e^{\frac{1}{2}\log(A)} = A^{\frac{1}{2}}$. We know that $(e^{\frac{1}{2}\log(A)})^2 = e^{\frac{1}{2}\log(A)}e^{\frac{1}{2}\log(A)} = e^{\log(A)} = A$. So $e^{\frac{1}{2}\log(A)} = \sqrt{A}$. Note that all eigenvalues of A are positive so all the eigenvalues of $\log(A)$ are real, which means the eigenvalues of $e^{\frac{1}{2}\log(A)}$ are all positive. So $e^{\frac{1}{2}\log(A)} = A^{\frac{1}{2}}$.

Note that $\frac{1}{2}\log(A)$ and $\frac{1}{2}\log(B)$ commute. We have

$$\begin{aligned} E(A, B) &= e^{\frac{1}{2}\log(A) + \frac{1}{2}\log(B)} \\ &= e^{\frac{1}{2}\log(A)}e^{\frac{1}{2}\log(B)} \\ &= A^{\frac{1}{2}}B^{\frac{1}{2}} \end{aligned}$$

\square

Problem 2.5

(2.27a) Note that $AA^{-1}A = A$. From definition, $A\#A = A$.

(2.27b)

$$\begin{aligned}
(A\#B)^{-1} &= (B(B^{-1}A)^{\frac{1}{2}})^{-1} \\
&= (B^{-1}A)^{-\frac{1}{2}}B^{-1} \\
&= ((B^{-1}A)^{-1})^{\frac{1}{2}}B^{-1} \\
&= (A^{-1}B)^{\frac{1}{2}}B^{-1} \\
&= A^{-1}\#B^{-1}
\end{aligned}$$

(2.27c)

$$\begin{aligned}
A\#B &= B(B^{-1}A)^{\frac{1}{2}} \\
&= AA^{-1}B(B^{-1}A)^{\frac{1}{2}} \\
&= A(B^{-1}A)^{-1}(B^{-1}A)^{\frac{1}{2}} \\
&= A(B^{-1}A)^{-\frac{1}{2}} \\
&= A(A^{-1}B)^{\frac{1}{2}} \\
&= B\#A
\end{aligned}$$

(2.27d) Use “ ≥ 0 ” to denote hermitian positive semidefinite. For any complex square matrix $C \geq 0$ and $D \geq 0$, for $\forall x \in \mathbb{C}^n$, $x^*CDCx = x^*C^*DCx = (Cx)^*D(Cx) \geq 0$. So $CDC \geq 0$.

Let $T = B^{-\frac{1}{2}}AB^{-\frac{1}{2}}$. Then $T \geq 0$. Let $T = P\Lambda P^*$. we have $(T^{\frac{1}{2}} - I)^2 = P(\Lambda^{\frac{1}{2}} - I)^2P^*$, which means $(T^{\frac{1}{2}} - I)^2 \geq 0$. Then we know $(T + I - 2T^{\frac{1}{2}}) \geq 0$. Simultaneously multiply hermitian positive definite $B^{\frac{1}{2}}$ from both sides, we get $(B^{\frac{1}{2}}TB^{\frac{1}{2}} + B - 2B^{\frac{1}{2}}T^{\frac{1}{2}}B^{\frac{1}{2}}) \geq 0$. I.e. $(A + B - 2A\#B) \geq 0$. \square

Problem 2.6

Let $X = (A^{-1}B)^{\frac{1}{2}}$ and the two eigenvalues of X be λ_1, λ_2 . We know $tr(X) = \lambda_1 + \lambda_2$, $det(X) = \lambda_1\lambda_2 = \frac{\beta}{\alpha}$. Then

$$\begin{aligned}
det(\alpha^{-1}A + \beta^{-1}B) &= det(A \cdot (\alpha^{-1}I + \beta^{-1}X^2)) \\
&= det(A) \cdot det(\alpha^{-1}I + \beta^{-1}X^2) \\
&= \alpha^2 \left(\frac{1}{\alpha} + \frac{\lambda_1^2}{\beta}\right) \left(\frac{1}{\alpha} + \frac{\lambda_2^2}{\beta}\right) \\
&= \frac{\alpha}{\beta} (\lambda_1 + \lambda_2)^2 \\
&= \frac{\alpha}{\beta} tr(X)^2
\end{aligned}$$

We know that X is the principal square root of $A^{-1}B$, which means $tr(X) \geq 0$. So

$$tr(X) = \sqrt{\frac{\beta}{\alpha} det(\alpha^{-1}A + \beta^{-1}B)}$$

From Cayley-Hamilton Theorem, we have $X^2 - tr(X)X + det(X)I = O$, i.e.

$$tr(X)(A^{-1}B)^{\frac{1}{2}} = A^{-1}B + det(X)I$$

Multiply A from the left, we get

$$tr(X)(A\#B) = det(X)A + B$$

Simplify this equation, we have

$$\begin{aligned} A \# B &= \sqrt{\frac{\alpha}{\beta \det(\alpha^{-1}A + \beta^{-1}B)}} B + \frac{\beta}{\alpha} \sqrt{\frac{\alpha}{\beta \det(\alpha^{-1}A + \beta^{-1}B)}} A \\ &= \frac{\sqrt{\alpha\beta}}{\sqrt{\det(\alpha^{-1}A + \beta^{-1}B)}} (\alpha^{-1}A + \beta^{-1}B) \end{aligned}$$

□

Problem 2.7

Use “ > 0 ” to denote hermitian positive definite. For $\forall x \in \mathbb{C}^n$, $x^* R B R^* x = (R^* x)^* B (R^* x) > 0$, so $R B R^* > 0$, thus $(R B R^*)^{\frac{1}{2}} > 0$. $x^* R^{-1} (R B R^*)^{\frac{1}{2}} R^{-*} x = (R^{-*} x)^* (R B R^*)^{\frac{1}{2}} (R^{-*} x) > 0$, then $R^{-1} (R B R^*)^{\frac{1}{2}} R^{-*} > 0$.

We know that the hermitian positive definite solution to $X A X = B$ is unique (proven in Page 45). So we verify that $X = R^{-1} (R B R^*)^{\frac{1}{2}} R^{-*}$ is a solution. We have

$$X A X = R^{-1} (R B R^*)^{\frac{1}{2}} R^{-*} R^* R R^{-1} (R B R^*)^{\frac{1}{2}} R^{-*} = B$$

So $X = R^{-1} (R B R^*)^{\frac{1}{2}} R^{-*}$ is the hermitian positive definite solution to $X A X = B$. We can compute the solution using the formula of X . □