

Lecture 8

①

Gaussian states:

- Previously, we considered the evolution generated by Hamiltonians that are linear or strictly quadratic in the quadrature operators.
- Now, we will consider the thermal states corresponding to quadratic Hamiltonians.

Consider a Hamiltonian of the form

$$\hat{H} = \frac{1}{2} \hat{r}^T H \hat{r} + \hat{r}^T \bar{r}'$$

where $\bar{r}' \in \mathbb{R}^{2n}$ & H is a positive definite $2n \times 2n$ real matrix

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A faithful Gaussian, ^{n-mode} state is

defined as
$$\frac{e^{-\beta \hat{H}}}{\text{Tr}[e^{-\beta \hat{H}}]} \quad \text{for } \beta > 0$$

faithful means that the state is positive definite (has full support)

Consider that

$$\begin{aligned} \hat{H}' &= \frac{1}{2} (\hat{r} - \bar{r})^T H (\hat{r} - \bar{r}) \\ &= \frac{1}{2} (\hat{r}^T H \hat{r} - 2 \bar{r}^T H \hat{r} + \|\bar{r}\|_2^2) \\ &= \frac{1}{2} \hat{r}^T H \hat{r} - \hat{r}^T H \bar{r} + \frac{1}{2} \|\bar{r}\|_2^2 \end{aligned}$$

Then ~~substituting~~ setting $\bar{r} = -H^{-1} \bar{r}'$,
we recover the original form of the Hamiltonian stated.

Also, when exponentiating \hat{H}' as

$e^{-\beta \hat{H}'}$, the constant term $\frac{1}{2} \|\bar{r}\|_2^2$ is eliminated after normalization

2a

It is natural to consider computing the mean vector, covariance matrix, & normalization for a faithful Gaussian state as a function of \bar{r} & H .

Let us start simple. Suppose a single-mode state w/ $H = \lambda \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ for $\lambda > 0$ & $\bar{r} = 0$.

Then the state is given by

$$\frac{e^{-\frac{1}{2} \hat{r}^T H \hat{r}}}{\text{Tr}[e^{-\frac{1}{2} \hat{r}^T H \hat{r}}]}$$

Consider that

$$\begin{aligned} \frac{1}{2} \hat{r}^T H \hat{r} &= \frac{1}{2} \begin{bmatrix} \hat{x} & \hat{p} \end{bmatrix} \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} \begin{bmatrix} \hat{x} \\ \hat{p} \end{bmatrix} \\ &= \frac{\lambda}{2} (\hat{x}^2 + \hat{p}^2) \end{aligned}$$

Now use that $\hat{n} = \hat{a}^\dagger \hat{a} = \left(\frac{\hat{x} - i\hat{p}}{\sqrt{2}} \right) \left(\frac{\hat{x} + i\hat{p}}{\sqrt{2}} \right)$

(2b)

$$= \frac{1}{2} \left[\hat{x}^2 + \hat{p}^2 + i [\hat{x}, \hat{p}] \right]$$

$$= \frac{1}{2} \left[\hat{x}^2 + \hat{p}^2 - 1 \right] \Rightarrow \frac{\hat{x}^2 + \hat{p}^2}{2} = \hat{n} + \frac{1}{2}$$

Then $\frac{1}{2} \hat{r}^T H \hat{r} = \lambda \left(\hat{n} + \frac{1}{2} \right)$

Now use that $\hat{n} = \sum_{n=0}^{\infty} n |n\rangle \langle n|$

to write $e^{-\frac{1}{2} \hat{r}^T H \hat{r}}$ as $\Rightarrow f\left(\hat{n} + \frac{1}{2}\right) = \sum_{n=0}^{\infty} f\left(n + \frac{1}{2}\right) |n\rangle \langle n|$

$$\Rightarrow e^{-\frac{1}{2} \hat{r}^T H \hat{r}} = \sum_{n=0}^{\infty} e^{-\lambda \left(n + \frac{1}{2}\right)} |n\rangle \langle n|$$

$$= e^{-\lambda/2} \sum_{n=0}^{\infty} e^{-\lambda n} |n\rangle \langle n|$$

$$\Rightarrow \text{Tr} \left[e^{-\frac{1}{2} \hat{r}^T H \hat{r}} \right] = e^{-\lambda/2} \sum_{n=0}^{\infty} e^{-\lambda n}$$

$$= e^{-\lambda/2} \frac{1}{1 - e^{-\lambda}}$$

$$= \frac{1}{e^{\lambda/2} - e^{-\lambda/2}}$$

$$\equiv z(\lambda)$$

(2c)

What about the mean vector?

Consider that $\langle n | \tilde{x} | n \rangle$

$$= \langle n | \frac{\hat{a} + \hat{a}^\dagger}{\sqrt{2}} | n \rangle$$

$$= \frac{1}{\sqrt{2}} \left[\langle n | \hat{a} | n \rangle + \langle n | \hat{a}^\dagger | n \rangle \right]$$

$$= \frac{1}{\sqrt{2}} \left[\langle n | \sqrt{n} | n-1 \rangle + \langle n | \sqrt{n+1} | n+1 \rangle \right]$$

$$= 0$$

Similarly, $\langle n | \tilde{p} | n \rangle = 0$

So any Fock-diagonal state has
mean ^{vector equal to} zero & thus

$$\text{Tr} \left[e^{-\frac{1}{2} \hat{r}^T H \hat{r}} \right] = 0 = \text{Tr} \left[e^{-\frac{1}{2} \hat{r}^T H \hat{r}} \hat{p} \right]$$

What about the ^{co-}variance matrix?

i.e. $\text{Tr} \left[\{ \hat{r}_i, \hat{r}_j \} e^{-\frac{1}{2} \hat{r}^T H \hat{r}} \right]$

can show that $\langle n | \hat{x} \hat{p} + \hat{p} \hat{x} | n \rangle = 0$

$$\langle n | 2\hat{x}^2 | n \rangle = \langle n | 2\hat{p}^2 | n \rangle = 2n + 1$$

so show

$$\text{Tr} \left[\frac{e^{-\frac{1}{2}\hat{r}^T H \hat{r}}}{z(\lambda)} \begin{Bmatrix} \hat{x} \\ \hat{p} \end{Bmatrix} \right] = 0$$

$$\neq 2 \text{Tr} \left[\hat{x}^2 \frac{e^{-\frac{1}{2}\hat{r}^T H \hat{r}}}{z(\lambda)} \right]$$

$$= \frac{2}{z(\lambda)} \text{Tr} \left[\hat{x}^2 \sum_{n=0}^{\infty} e^{-\lambda(n+\frac{1}{2})} |n\rangle\langle n| \right]$$

$$= \frac{2}{z(\lambda)} \sum_{n=0}^{\infty} e^{-\lambda(n+\frac{1}{2})} \langle n | \hat{x}^2 | n \rangle$$

$$= \frac{2}{z(\lambda)} \sum_{n=0}^{\infty} e^{-\lambda(n+\frac{1}{2})} (n+\frac{1}{2})$$

$$= 1 + 2 \frac{e^{-\lambda/2}}{z(\lambda)} \sum_{n=0}^{\infty} e^{-\lambda n} n$$

$$= 1 + 2 \frac{e^{-\lambda/2}}{z(\lambda)} \left[-\frac{d}{d\lambda} \left(\sum_{n=0}^{\infty} e^{-\lambda n} \right) \right]$$

$$= 1 + 2 \frac{e^{-\lambda/2}}{z(\lambda)} \left[\frac{d}{d\lambda} \left(\frac{1}{1-e^{-\lambda}} \right) \right]$$

$$= \coth\left(\frac{\lambda}{2}\right) \equiv v(\lambda) > 1 \quad \text{for } \lambda > 0$$

where $\coth(x) = \frac{e^x + e^{-x}}{e^x - e^{-x}}$

Similarly, $2 \text{Tr} \left[\rho^{-1/2} \frac{e^{-\frac{1}{2} \hat{r}^T H \hat{r}}}{z(\lambda)} \right] = 2 \coth\left(\frac{\lambda}{2}\right) \quad (2e)$

So a single-mode Hamiltonian ~~is~~

$$H = \begin{bmatrix} \lambda & 0 \\ 0 & \lambda \end{bmatrix} \quad \text{for } \lambda > 0$$

has mean vector equal to zero,

covariance matrix given by $\begin{bmatrix} v(\lambda) & 0 \\ 0 & v(\lambda) \end{bmatrix}$

where $v(\lambda) = \coth\left(\frac{\lambda}{2}\right) > 1$

↓ normalization

$$\text{Tr} \left[e^{-\frac{1}{2} \hat{r}^T H \hat{r}} \right] = z(\lambda) = \frac{1}{e^{\lambda/2} - e^{-\lambda/2}}$$

If we had instead started by specifying

the covariance matrix as $\sigma = \begin{bmatrix} v & 0 \\ 0 & v \end{bmatrix}$

for v satisfying $\begin{bmatrix} v & 0 \\ 0 & v \end{bmatrix} + i \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix} \succcurlyeq 0$

$v > 0$

$\Leftrightarrow \begin{matrix} v+1 \succcurlyeq 0, \\ v-1 \succcurlyeq 0 \end{matrix} \Rightarrow v \succcurlyeq 1$

then the Hamiltonian matrix elements are given by the inverse operation

$$\lambda(v) = 2 \operatorname{arccoth}(v) = 2 \operatorname{arccoth}(v) \quad |x| > 1$$

where $\operatorname{arccoth}(x) = \frac{1}{2} \ln \left(\frac{x+1}{x-1} \right)$

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then the Hamiltonian ^{operator} is

$$\frac{1}{2} \hat{r}^T H \hat{r} = \frac{1}{2} \sum_{j=1}^n \lambda_j (\hat{x}_j^2 + \hat{p}_j^2)$$

$$\downarrow \text{so } e^{-\frac{1}{2} \hat{r}^T H \hat{r}} = e^{-\frac{1}{2} \sum_{j=1}^n \lambda_j (\hat{x}_j^2 + \hat{p}_j^2)}$$

$$= \bigotimes_{j=1}^n e^{-\frac{\lambda_j}{2} (\hat{x}_j^2 + \hat{p}_j^2)}$$

So arguments from last ~~the~~ ^{calculation} apply:

mean vector is equal to zero

Normalization is given by

$$\begin{aligned} & \text{Tr} \left[\bigotimes_{j=1}^n e^{-\frac{\lambda_j}{2} (\hat{x}_j^2 + \hat{p}_j^2)} \right] \\ &= \prod_{j=1}^n \text{Tr} \left[e^{-\frac{\lambda_j}{2} (\hat{x}_j^2 + \hat{p}_j^2)} \right] \\ &= \prod_{j=1}^n z(\lambda_j) \end{aligned}$$

covariance matrix is diagonal matrix w/

$$\sigma = \bigoplus_{j=1}^n v(\lambda_j) \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \quad \text{where } v(\lambda_j) = \coth \frac{\lambda_j}{2} > 1$$

2h

If covariance matrix elements are given as $\bigoplus_{j=1}^n v_j \begin{bmatrix} 1 & 0 \\ 0 & i \end{bmatrix} = \sigma$ w/ $v_j > 1$

\Rightarrow Hamiltonian $\bigoplus_{j=1}^n h(v_j) \begin{bmatrix} 1 & 0 \\ 0 & i \end{bmatrix} \otimes -h(v)$
Then normalization is $= 2 \operatorname{arccoth}(v) > 0$

$$\begin{aligned} \prod_{j=1}^n z(h(v_j)) &= \prod_{j=1}^n \frac{1}{2} \sqrt{v_j^2 - 1} \\ &= \prod_{j=1}^n \sqrt{\operatorname{Det} \left(\frac{\sigma_j + i\Omega_j}{2} \right)} \\ &= \sqrt{\prod_{j=1}^n \operatorname{Det} \left(\frac{\sigma_j + i\Omega_j}{2} \right)} \\ &= \sqrt{\operatorname{Det} \left(\frac{\sigma + i\Omega}{2} \right)} \end{aligned}$$

because

$$\sigma + i\Omega = \bigoplus_{j=1}^n \sigma_j + i\Omega_j$$

$$\text{where } \sigma_j = v_j \begin{bmatrix} 1 & 0 \\ 0 & i \end{bmatrix}$$

we used the fact that

$$\operatorname{Det}(A \oplus B) = \operatorname{Det}(A) \cdot \operatorname{Det}(B)$$

(2i)

So for multimode states

w/ Hamiltonian matrix $H = \bigoplus_{j=1}^n \hbar_j \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

the state $\frac{e^{-\frac{1}{2}\hat{r}^T H \hat{r}}}{\text{Tr}[\dots]}$ has mean vector equal to zero, covariance matrix

$$\sigma = \bigoplus_{j=1}^n v(\hbar_j) \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

w/ $v(\hbar_j) = \coth\left(\frac{\hbar_j}{2}\right)$

+ normalization

$$\prod_{j=1}^n z(\hbar_j), \text{ which is equal to}$$

$$\sqrt{\text{Det}\left(\frac{\sigma + i\Omega}{2}\right)}$$

Suppose now that we take such a

diagonal Hamiltonian H &

act on it by congruence w/

a symplectic matrix S , to produce a new Ham. matrix

$$\Rightarrow H' = \cancel{S^T H S} = S^T H S$$

By earlier results,

where $S = e^{\Omega A}$
for symmetric real A

(2)

As shown previously

$$S_{\hat{r}} = e^{\Lambda A_{\hat{r}}} = e^{\frac{i}{2} \hat{r}^T A \hat{r}} \hat{r} e^{-\frac{i}{2} \hat{r}^T A \hat{r}}$$

for real symmetric A

$$\begin{aligned} \Rightarrow S_{\hat{r}}^{-1} &= e^{-\Lambda A_{\hat{r}}} = e^{\frac{i}{2} \hat{r}^T (-A) \hat{r}} \hat{r} e^{-\frac{i}{2} \hat{r}^T (-A) \hat{r}} \\ &= e^{-\frac{i}{2} \hat{r}^T A \hat{r}} \hat{r} e^{\frac{i}{2} \hat{r}^T A \hat{r}} \end{aligned}$$

$$\Rightarrow \frac{1}{2} \hat{r}^T S^T H S \hat{r} = \frac{1}{2} (S \hat{r})^T H S \hat{r}$$

$$= \frac{1}{2} \left(e^{\frac{i}{2} \hat{r}^T A \hat{r}} \hat{r} e^{-\frac{i}{2} \hat{r}^T A \hat{r}} \right)^T H \left(e^{\frac{i}{2} \hat{r}^T A \hat{r}} \hat{r} e^{-\frac{i}{2} \hat{r}^T A \hat{r}} \right)$$

$$= e^{\frac{i}{2} \hat{r}^T A \hat{r}} \frac{1}{2} \hat{r}^T H \hat{r} e^{-\frac{i}{2} \hat{r}^T A \hat{r}}$$

$$\Rightarrow e^{-\frac{1}{2} \hat{r}^T S^T H S \hat{r}}$$

$$= e^{\frac{i}{2} \hat{r}^T A \hat{r}} e^{-\frac{1}{2} \hat{r}^T H \hat{r}} e^{-\frac{i}{2} \hat{r}^T A \hat{r}}$$

$$\Rightarrow \text{mean vector of } \frac{e^{-\frac{1}{2} \hat{r}^T S^T H S \hat{r}}}{\text{Tr}[\cdot]}$$

$$= \text{Tr} \left[\hat{r} e^{\frac{i}{2} \hat{r}^T A \hat{r}} \frac{e^{-\frac{1}{2} \hat{r}^T H \hat{r}} e^{-\frac{i}{2} \hat{r}^T A \hat{r}}}{\text{Tr}[\cdot]} \right]$$

(2k)

$$= \text{Tr} \left[e^{-\frac{i}{2} \hat{r}^T A \hat{r}} \hat{r} e^{\frac{i}{2} \hat{r}^T A \hat{r}} \frac{e^{-\frac{1}{2} \hat{r}^T H \hat{r}}}{\text{Tr}[\cdot]} \right]$$

$$= \text{Tr} \left[S^{-1} \hat{r} e^{-\frac{1}{2} \hat{r}^T H \hat{r}} \frac{1}{\text{Tr}[\cdot]} \right]$$

$$= S^{-1} \text{Tr} \left[\hat{r} e^{-\frac{1}{2} \hat{r}^T H \hat{r}} \frac{1}{\text{Tr}[\cdot]} \right] = S^{-1} \cdot 0 = 0$$

What is covariance matrix?

given by (since zero mean)

$$\text{Tr} \left[\left\{ \hat{r}, \hat{r}^T \right\} e^{\frac{i}{2} \hat{r}^T A \hat{r}} \frac{e^{-\frac{1}{2} \hat{r}^T H \hat{r}}}{\text{Tr}[\cdot]} e^{-\frac{i}{2} \hat{r}^T A \hat{r}} \right]$$

$$= \text{Tr} \left[e^{-\frac{i}{2} \hat{r}^T A \hat{r}} \left\{ \hat{r}, \hat{r}^T \right\} e^{\frac{i}{2} \hat{r}^T A \hat{r}} \frac{e^{-\frac{1}{2} \hat{r}^T H \hat{r}}}{\text{Tr}[\cdot]} \right]$$

$$= \text{Tr} \left[\left\{ S^{-1} \hat{r}, (S^{-1} \hat{r})^T \right\} e^{-\frac{1}{2} \hat{r}^T H \hat{r}} \frac{1}{\text{Tr}[\cdot]} \right]$$

$$= S^{-1} \text{Tr} \left[\left\{ \hat{r}, \hat{r}^T \right\} e^{-\frac{1}{2} \hat{r}^T H \hat{r}} \frac{1}{\text{Tr}[\cdot]} \right] S^{-T}$$

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$$= S^{-1} \sigma S^{-T} = \sigma'$$

$$\Rightarrow \sigma' = S^{-1} \left(\bigoplus_{j=1}^n \cosh\left(\frac{f_j}{2}\right) \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \right) S^{-T}$$

What about the normalization factor?

$$\text{Tr} \left[e^{-\frac{1}{2} \hat{r}^T H' \hat{r}} \right]$$

$$= \text{Tr} \left[e^{\frac{i}{2} \hat{r}^T A \hat{r}} e^{-\frac{1}{2} \hat{r}^T H' \hat{r}} e^{-\frac{i}{2} \hat{r}^T A \hat{r}} \right]$$

$$= \text{Tr} \left[e^{-\frac{1}{2} \hat{r}^T H' \hat{r}} \right] = \sqrt{\text{Det} \left(\frac{\sigma + i\Omega}{2} \right)}$$

Using the fact that

$$\text{Det}(S) = 1 = \text{Det}(S^{-1})$$

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

symplectic S , we have that

$$\begin{aligned} \sqrt{\text{Det} \left(\frac{\sigma + i\Omega}{2} \right)} &= \sqrt{\text{Det}(S^{-1}) \text{Det} \left(\frac{\sigma + i\Omega}{2} \right) \text{Det}(S^{-T})} \\ &= \sqrt{\text{Det} \left(S^{-1} \left(\frac{\sigma + i\Omega}{2} \right) S^{-T} \right)} = \sqrt{\text{Det} \left(\frac{\sigma' + i\Omega}{2} \right)} \end{aligned}$$

Z_m

$$= \sqrt{\text{Det} \left(\frac{\sigma' + i\Lambda}{2} \right)}$$

where we used that $\sum_A \Lambda_A \Lambda_A^T = \Lambda$

Now suppose that we act on new state w/ displacement operator

$$\hat{D}_{-\bar{r}} e^{-\frac{1}{2} \hat{r}^T H' \hat{r}} \hat{D}_{+\bar{r}}$$

where $\hat{D}_{\bar{r}} = \exp(i \bar{r}^T \Lambda \hat{r})$

This is equal to

$$e^{-\frac{1}{2} [\hat{D}_{-\bar{r}} \hat{r}^T H' \hat{r} \hat{D}_{\bar{r}}]}$$

↳ now write $\hat{r}^T H' \hat{r} = \sum_{jk} \hat{r}_j H'_{jk} \hat{r}_k$

↳ see that $\hat{D}_{-\bar{r}} \hat{r}^T H' \hat{r} \hat{D}_{\bar{r}} =$

$$\begin{aligned} & \sum_{jk} \hat{D}_{-\bar{r}} \hat{r}_j \hat{D}_{\bar{r}} H'_{jk} \hat{D}_{-\bar{r}} \hat{r}_k \hat{D}_{\bar{r}} \\ &= \sum_{jk} (\hat{r}_j - \bar{r}_j) H'_{jk} (\hat{r}_k - \bar{r}_k) \end{aligned}$$

(2n)

$$\Rightarrow e^{-\frac{1}{2}(\hat{r}-\bar{r})^T H' (\hat{r}-\bar{r})}$$

under this change,

mean vector translates from zero to \bar{r} .

covariance matrix σ' unchanged,

because it is invariant under changes of the mean vector.

Normalization unchanged.

So we can write this faithful

Gaussian state as

$$\frac{e^{-\frac{1}{2}(\hat{r}-\bar{r})^T H' (\hat{r}-\bar{r})}}{\sqrt{\text{Det}\left(\frac{\sigma' + i\Omega}{2}\right)}}$$

} notice similarity
to classical
multivariate
Gaussian
density.

$$\text{where } H' = S^T \bigoplus_{j=1}^n \lambda_j \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} S$$

$$\sigma' = S^{-1} \bigoplus_{j=1}^n \coth\left(\frac{\lambda_j}{2}\right) \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} S^{-T}$$

It turns out that this is
the most general form!!

2na

By every thing that we've done,
it follows that

$$\frac{e^{-\frac{1}{2}(\hat{r}-\bar{r})^T H'(\hat{r}-\bar{r})}}{\sqrt{\text{Det}\left(\frac{\sigma+i\ell}{2}\right)}} = \frac{\hat{D}_{-\bar{r}} \int_A \hat{D}_{\bar{r}} e^{-\frac{1}{2}\hat{r}^T H \hat{r}} \int_A \hat{D}_{\bar{r}}}{\text{Tr}\left[e^{-\frac{1}{2}\hat{r}^T H \hat{r}}\right]}$$

where $H = \bigoplus_{j=1}^n \lambda_j \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$

$$\hat{S}_A = e^{\frac{i}{2}\hat{r}^T A \hat{r}}$$

for $\lambda_j > 0$

$$\begin{aligned} \int \text{Tr}\left[e^{-\frac{1}{2}\hat{r}^T H \hat{r}}\right] &= \sqrt{\text{Det}\left(\frac{\sigma+i\ell}{2}\right)} \\ &= \sqrt{\text{Det}\left(\frac{\sigma+i\ell}{2}\right)} \end{aligned}$$

$$\text{w/ } \sigma = \bigoplus_{j=1}^n v(\lambda_j) \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

where $v(\lambda) = \coth\left(\frac{\lambda}{2}\right)$

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Why \Rightarrow the determinant of a symplectic matrix S ~~is~~ equal to ± 1 ? ($\det(S) = 1$)

Consider that $S \Omega S^T = \Omega$

$$\Rightarrow \det(S \Omega S^T) = \det(\Omega)$$

Since $\det(\Omega) = 1$

$$\Rightarrow \det(S) \det(\Omega) \det(S^T) = \det(\Omega) = 1$$

$$\Rightarrow \det(S) \det(S^T) = 1$$

$$\Rightarrow \{\det(S)\}^2 = 1$$

$$\text{So } \det(S) = +1 \text{ or } -1$$

Now eliminate possibility that $\det(S) = -1$

Since any symplectic S is invertible (thus full rank) it follows that

$S^T S$ is symmetric positive definite

\Rightarrow eigenvalues of $S^T S + I$ are > 1 .

$$\begin{aligned} \text{Then } S^T S + I &= S^T (S + S^{-T}) \\ &= S^T (S + \Omega S \Omega^T) \end{aligned}$$

(2p)

$$\text{b/c } S \Lambda S^T = \Lambda \Rightarrow$$

$$S \Lambda S^T \Lambda^T = \Lambda \Lambda^T = I$$

$$\Rightarrow S^{-1} = \Lambda S^T \Lambda^T$$

$$\Rightarrow S^{-T} = \Lambda S \Lambda^T$$

Consider that $\Lambda = \bigoplus_{j=1}^n \Lambda_j = I_n \otimes \Lambda_1$

Now write $S = \sum_{j,k \in \{0,1\}} S_{j,k} \otimes |j\rangle\langle k|$ $\Lambda_1 = \begin{bmatrix} 0 & 1 \\ -1 & 0 \end{bmatrix}$

$$\Rightarrow S + \Lambda S \Lambda^T$$

$$= \sum_{j,k} S_{j,k} \otimes |j\rangle\langle k| + (I_n \otimes \Lambda_1) (S_{j,k} \otimes |j\rangle\langle k|) (I_n \otimes \Lambda_1^T)$$

$$= \sum_{j,k} S_{j,k} \otimes [|j\rangle\langle k| + \Lambda_1 |j\rangle\langle k| \Lambda_1^T]$$

$$\Lambda_1 |0\rangle = -|1\rangle$$

$$\Lambda_1 |1\rangle = |0\rangle$$

$$\Rightarrow |j\rangle\langle k| + \Lambda_1 |j\rangle\langle k| \Lambda_1^T = |j\rangle\langle k| + (-1)^{j \oplus 1} |j \oplus 1\rangle\langle k \oplus 1|$$

$$= |j\rangle\langle k| + (-1)^{j \oplus k} |j \oplus 1\rangle\langle k \oplus 1|$$

2a

$$\Rightarrow = (S_{00} + S_{11}) \otimes |0\rangle\langle 0| + (S_{01} - S_{10}) \otimes |0\rangle\langle 1| \\ + (-S_{01} + S_{10}) \otimes |1\rangle\langle 0| + (S_{00} + S_{11}) \otimes |1\rangle\langle 1|$$

Define $C = S_{00} + S_{11}$ } these are
 $D = S_{01} - S_{10}$ } real matrices

$$\Rightarrow S + \Lambda S \Lambda^T = C \otimes |0\rangle\langle 0| + D \otimes |0\rangle\langle 1| \\ - D \otimes |1\rangle\langle 0| + C \otimes |1\rangle\langle 1|$$

$$= (\mathbf{I} \otimes \underline{u}) \left([C + iD] \otimes |0\rangle\langle 0| + [C - iD] \otimes |1\rangle\langle 1| \right) (\mathbf{I} \otimes \underline{u}^T)$$

where $\underline{u} = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 & 1 \\ i & -i \end{bmatrix}$

Then

$$0 < 1 < \det(S^T S + \mathbf{I})$$

$$= \det(S^T (S + \Lambda S \Lambda^T))$$

$$= \det(S^T) \det(S + \Lambda S \Lambda^T)$$

$$= \det(S) \det(\mathbf{I} \otimes \underline{u}) \det(C + iD) \det(C - iD)$$

$$= \det(S) \det(C + iD) \det(\overline{C + iD}) \det(\mathbf{I} \otimes \underline{u}^T)$$

$$= \det(S) \det(C + iD) \overline{\det(C + iD)}$$

$$= \det(s) |\det(C+iD)|^2$$

(2r)

Since $\det(s) |\det(C+iD)|^2 > 0$

must be the case that

$$|\det(C+iD)|^2 > 0$$

can divide through &

conclude that $\det(s) > 0$

$$\Rightarrow \det(s) = 1$$

by elimination.