

Project 3 (Group member: Jianqing Li (alone))

$$\begin{aligned}
 Q1) R_k^{BC} &\leq I(\underline{S}_k: y_k | \underline{S}_1, \dots, \underline{S}_{k-1}) \\
 &= h(y_k | \underline{S}_1, \dots, \underline{S}_{k-1}) - h(y_k | \underline{S}_1, \dots, \underline{S}_k) \\
 &= \log_2 \det(\Pi(\underline{I}_M + H_k(\sum_{i=k}^K \underline{S}_i) H_k^H)) - \\
 &\quad \log_2 \det(\Pi(\underline{I}_M + H_k(\sum_{i=k+1}^K \underline{S}_i) H_k^H)) \\
 &= \log_2 \det(\underline{I}_M + (\underline{I}_M + H_k(\sum_{i=k+1}^K \underline{S}_i) H_k^H)^{-1} H_k \underline{S}_k H_k^H) \\
 &= \log_2 \det(\underline{I}_N + H_k^H (\underline{I}_M + H_k(\sum_{i=k+1}^K \underline{S}_i) H_k^H)^{-1} H_k \underline{S}_k)
 \end{aligned}$$

$$\begin{aligned}
 Q2) R_k^{MAC} &\leq I(\underline{x}_k, y | \underline{x}_{k+1}, \dots, \underline{x}_K) \\
 &= h(y | \underline{x}_{k+1}, \dots, \underline{x}_K) - h(y | \underline{x}_k, \underline{x}_{k+1}, \dots, \underline{x}_K) \\
 &= \log_2 \det(\Pi(\underline{I}_N + \sum_{i=1}^k H_i^H \underline{Q}_i H_i)) - \log_2 \det(\Pi(\underline{I}_N + \sum_{i=1}^{k-1} H_i^H \underline{Q}_i H_i)) \\
 &= \log_2 \det(\underline{I}_N + (\underline{I}_N + \sum_{i=1}^{k-1} H_i^H \underline{Q}_i H_i)^{-1} H_k^H \underline{Q}_k H_k) \\
 &= \log_2 \det(\underline{I}_M + H_k (\underline{I}_N + \sum_{i=1}^{k-1} H_i^H \underline{Q}_i H_i)^{-1} H_k^H \underline{Q}_k)
 \end{aligned}$$

$$Q3) S = V U^H Q U V^H = V U^H [U \ U'] \begin{bmatrix} Q_{11} & Q_{12} \\ Q_{12}^H & Q_{22} \end{bmatrix} \begin{bmatrix} U^H \\ U'^H \end{bmatrix} U V^H$$

For  $r=N$  since  $[U \ U']$  and  $Q$  are unitary,  $U$  is subunitary

we have  $U U' = 0$   $U'^H U = 0$   $U^H U = \underline{I}_r$

$$\begin{aligned}
 \text{Thus, } S &= V \begin{bmatrix} \underline{I}_r & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} Q_{11} & Q_{12} \\ Q_{12}^H & Q_{22} \end{bmatrix} \begin{bmatrix} \underline{I}_r \\ 0 \end{bmatrix} V^H \\
 &= V Q_{11} V^H
 \end{aligned}$$

Finally, we have  $\text{tr}(S) = \text{tr}(Q_{11} V^H V) = \text{tr}(Q_{11})$

$$\text{tr}(Q) = \text{tr}([U \ U'] \begin{bmatrix} Q_{11} & Q_{12} \\ Q_{12}^H & Q_{22} \end{bmatrix} \begin{bmatrix} U^H \\ U'^H \end{bmatrix}) = \text{tr}(\begin{bmatrix} Q_{11} & Q_{12} \\ Q_{12}^H & Q_{22} \end{bmatrix})$$

$$= \text{tr}(Q_{11}) + \text{tr}(Q_{22})$$

since  $Q_{11}, Q_{22}$  are positive semi-definite

$$\text{We have } \text{tr}(S) = \text{tr}(Q_{11}) \leq \text{tr}(Q_{11}) + \text{tr}(Q_{22}) = \text{tr}(Q)$$

$$\Rightarrow \text{tr}(S) \leq \text{tr}(Q)$$

$$\begin{aligned}
 \text{For } r=M \quad S &= V U^H Q U V^H \quad \text{tr}(S) = \text{tr}(V U^H Q U V^H) = \text{tr}(U^H Q U V^H V) \\
 &\Rightarrow U \text{ is not only subunitary, } = \text{tr}(U^H Q U) = \text{tr}(U U^H Q) = \text{tr}(Q) \\
 &\text{but also unitary} \quad \Rightarrow \text{tr}(S) = \text{tr}(Q)
 \end{aligned}$$



$$Q4) R_k^{MAC} = \log_2 \det(\underline{I} + \underbrace{\underline{X}_k^{-\frac{1}{2}} \underline{H}_k^H \underline{F}_k^{-\frac{1}{2}}}_{\underline{H}_{k,eff}^H} \underbrace{\underline{F}_k^{\frac{1}{2}} \underline{Q}_k \underline{F}_k^{\frac{H}{2}}}_{\underline{Q}_{k,eff}} \underbrace{\underline{F}_k^{-\frac{H}{2}} \underline{H}_k \underline{X}_k^{-\frac{H}{2}}}_{\underline{H}_{k,eff}})$$

$$= \log_2 \det(\underline{I} + \underline{H}_{k,eff}^H \underline{Q}_{k,eff} \underline{H}_{k,eff})$$

$$R_k^{BC} = \log_2 \det(\underline{I} + \underline{F}_k^{-\frac{H}{2}} \underline{H}_k \underline{X}_k^{-\frac{1}{2}} \underline{Q}_{k,eff} \underline{X}_k^{-\frac{1}{2}} \underline{H}_k^H \underline{F}_k^{-\frac{1}{2}})$$

$$= \log_2 \det(\underline{I} + \underline{H}_{k,eff} \underline{Q}_{k,eff} \underline{H}_{k,eff}^H) \text{ with } \underline{Q}_{k,eff} = \underline{X}_k^{-\frac{H}{2}} \underline{Q}_k \underline{X}_k^{-\frac{1}{2}}$$

$$R_k^{MAC} = \log_2 \det(\underline{I} + \underline{V}_k \underline{\Sigma}_k \underline{U}_k^H \underline{Q}_{k,eff} \underline{U}_k \underline{\Sigma}_k \underline{V}_k^H)$$

$$= \log_2 \det(\underline{I} + \underline{\Sigma}_k \underline{U}_k^H \underline{Q}_{k,eff} \underline{U}_k \underline{\Sigma}_k \underline{V}_k^H \underline{V}_k)$$

$$= \log_2 \det(\underline{I} + \underline{\Sigma}_k \underline{U}_k^H \underline{Q}_{k,eff} \underline{U}_k \underline{\Sigma}_k)$$

$$R_k^{BC} = \log_2 \det(\underline{I} + \underline{U}_k \underline{\Sigma}_k \underline{V}_k^H \underline{Q}_{k,eff} \underline{V}_k \underline{\Sigma}_k \underline{U}_k^H)$$

$$= \log_2 \det(\underline{I} + \underline{\Sigma}_k \underline{V}_k^H \underline{Q}_{k,eff} \underline{V}_k \underline{\Sigma}_k)$$

$$\underline{Q}_k = \underline{X}_k^{-\frac{H}{2}} \underline{V}_k \underline{U}_k^H \underline{Q}_{k,eff} \underline{U}_k \underline{V}_k^H \underline{X}_k^{-\frac{1}{2}}$$

↓

$$\underline{Q}_{k,eff} = \underline{X}_k^{\frac{H}{2}} \underline{Q}_k \underline{X}_k^{\frac{1}{2}} = \underline{V}_k \underline{U}_k^H \underline{Q}_{k,eff} \underline{U}_k \underline{V}_k^H$$

$$\underline{R}_k^{BC} = \log_2 \det(\underline{I} + \underline{\Sigma}_k \boxed{\underline{V}_k^H \underline{V}_k} \underline{U}_k^H \underline{Q}_{k,eff} \underline{U}_k \boxed{\underline{V}_k^H \underline{V}_k} \underline{\Sigma}_k)$$

$$= \log_2 \det(\underline{I} + \underline{\Sigma}_k \underline{U}_k^H \underline{Q}_{k,eff} \underline{U}_k \underline{\Sigma}_k)$$

Q5)

$$\text{Order} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}:$$

$$\text{tr}(S_1) = 5.3936$$

$$\text{tr}(S_2) = 4.6064$$

$$R_{BC} = R_{MAC} =$$

$$\begin{bmatrix} 5.1943 \\ 2.4790 \end{bmatrix}$$

$$\text{order} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}:$$

$$\text{tr}(S_1) = 3.4113$$

$$\text{tr}(S_2) = 6.5887$$

$$R_{BC} = R_{MAC} =$$

$$\begin{bmatrix} 4.1384 \\ 3.5349 \end{bmatrix}$$



Q6)

① sum power constraint

$$\text{tr}(\underline{Q}_1) \leq P_{tx} \quad \text{tr}(\underline{Q}_2) \leq P_{tx}$$

$$\alpha \text{tr}(\underline{Q}_1) + (1-\alpha) \text{tr}(\underline{Q}_2) \leq \alpha P_{tx} + (1-\alpha) P_{tx} = P_{tx}$$

$$\alpha \underline{Q}_1 + (1-\alpha) \underline{Q}_2 \preceq 0$$

② positive semidefiniteness

$$\underline{Q}_1 \succeq 0 \iff x^H \underline{Q}_1 x \geq 0 \quad \forall x$$

$$\underline{Q}_2 \succeq 0 \iff x^H \underline{Q}_2 x \geq 0 \quad \forall x$$

$$x^H (\alpha \underline{Q}_1 + (1-\alpha) \underline{Q}_2) x \stackrel{\text{linearity}}{=} \alpha x^H \underline{Q}_1 x + (1-\alpha) x^H \underline{Q}_2 x \geq 0$$

$$\alpha \underline{Q}_1 + (1-\alpha) \underline{Q}_2 \succeq 0$$

Q7)

$$P_{tx} = 10 \quad \underline{Q}[1] = \begin{bmatrix} 3.0273 & 0.4629 \\ 0.4629 & 3.6667 \end{bmatrix} \quad \underline{Q}[2] = \begin{bmatrix} 3.2078 & 0.5613 \\ 0.5613 & 0.0982 \end{bmatrix}$$

$$C_{sum} = \cancel{10.2623} \quad 10.3623$$

Q8)

$$\text{order} = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$S[1] = \begin{bmatrix} 2.4443 & 1.1883 & -1.5921 \\ 1.1883 & 2.8745 & 0.7750 \\ -1.5921 & 0.7750 & 2.0817 \end{bmatrix}$$

$$\text{tr}(S[1]) = 7.4005$$

$$S[2] = \begin{bmatrix} 1.3643 & -0.8969 & 0.9385 \\ -0.8969 & 0.5896 & -0.6169 \\ 0.9385 & -0.6169 & 0.6456 \end{bmatrix}$$

$$\text{tr}(S[2]) = 2.5995$$

$$\text{tr}(S[1]) + \text{tr}(S[2]) = 10$$

$$\text{order} = \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

$$S[1] = \begin{bmatrix} 0.7379 & 1.3966 & 0.3470 \\ 1.3966 & 3.1440 & -0.3567 \\ 0.3470 & -0.3567 & 2.2157 \end{bmatrix}$$

$$\text{tr}(S[1]) = 6.0976$$

$$S[2] = \begin{bmatrix} 2.8995 & -1.2306 & -1.1805 \\ -1.2306 & 0.5223 & 0.5010 \\ -1.1805 & 0.5010 & 0.4806 \end{bmatrix}$$

$$\text{tr}(S[2]) = 3.9024$$

$$\text{tr}(S[1]) + \text{tr}(S[2]) = 10$$



Q9) 24.3939

25dB: ~~24.3939~~ bits

29.3505

30dB: ~~29.3505~~ bits

slope:  $(29.3505 - 24.3939) / (30\text{dB} - 25\text{dB}) = 0.99132$  bits/dB

Q10)

① User 1 always has larger rate than user 2, ~~but in case~~

② Depending on the encoding order, the relative difference between user 1 and user 2 can be changed.

⇒ If user 1 is ~~encoded~~ first, the difference is larger.

⇒ If user 2 is ~~encoded~~ first, the difference is smaller.

③ The two decoding ~~order~~ <sup>encoded</sup> have larger rate difference at high SNR region.

Q11)

$$\max_{\mathbf{I}_k} \log_2 \det(\mathbf{I}_M + \underbrace{\mathbf{V}_k^H \mathbf{H}_k^H}_{\hat{\mathbf{H}}_k^H} \underbrace{\mathbf{C}_{\mathbf{H}_k}^{-1}}_{\hat{\mathbf{C}}_{\mathbf{H}_k}^{-1}} \mathbf{H}_k \mathbf{V}_k \mathbf{I}_k \mathbf{I}_k^H) \quad \text{s.t.} \quad \text{tr}(\mathbf{I}_k \mathbf{I}_k^H) \leq \frac{P_{\text{TX}}}{K}$$

Optimal  $\mathbf{I}_k$ :

$$\mathbf{I}_k \mathbf{I}_k^H = \mathbf{V} \mathbf{\bar{I}} \mathbf{V}^H \quad \hat{\mathbf{H}}_k^H \hat{\mathbf{C}}_{\mathbf{H}_k}^{-1} \hat{\mathbf{H}}_k = \mathbf{U} \mathbf{\bar{I}} \mathbf{U}^H$$

For  $\mathbf{I}_k \mathbf{I}_k^H$ :  $\mathbf{V} = \mathbf{U}$ ,  $\mathbf{\bar{I}}$  is found by waterfilling

$\mathbf{I}_k$  can be constructed by  $\mathbf{I}_k = \mathbf{U} \mathbf{\bar{I}}^{\frac{1}{2}}$



Q12)

$$\max_{\tilde{\mathbf{I}}} \log_2 \det(\tilde{\mathbf{I}}_{MK} + \underbrace{\tilde{\mathbf{V}}^H \tilde{\mathbf{H}}^H}_{\tilde{\mathbf{H}}^H} \tilde{\mathbf{C}}^{-1} \underbrace{\tilde{\mathbf{H}} \tilde{\mathbf{V}}}_{\tilde{\mathbf{H}}} \tilde{\mathbf{I}} \tilde{\mathbf{I}}^H) \quad \text{s.t. } \text{tr}(\tilde{\mathbf{I}} \tilde{\mathbf{I}}^H) \leq P_{Tx}$$

$\tilde{\mathbf{I}}$  is block diagonal

$\tilde{\mathbf{H}}$  is block diagonal,  $\tilde{\mathbf{C}}$  is block diagonal,  $\tilde{\mathbf{I}}_{MK}$  is block diagonal

The inverse of a blockdiagonal matrix is block diagonal, since

$$\begin{bmatrix} \tilde{\mathbf{A}} & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\mathbf{B}} \end{bmatrix} * \begin{bmatrix} \tilde{\mathbf{A}}^{-1} & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\mathbf{B}}^{-1} \end{bmatrix} = \begin{bmatrix} \tilde{\mathbf{I}} & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\mathbf{I}} \end{bmatrix} = \tilde{\mathbf{I}}$$

The product of blockdiagonal matrices is block diagonal, since

$$\begin{bmatrix} \tilde{\mathbf{A}} & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\mathbf{B}} \end{bmatrix} \begin{bmatrix} \tilde{\mathbf{C}} & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\mathbf{D}} \end{bmatrix} = \begin{bmatrix} \tilde{\mathbf{A}}\tilde{\mathbf{C}} & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\mathbf{B}}\tilde{\mathbf{D}} \end{bmatrix}$$

Thus,  $\tilde{\mathbf{H}}^H \tilde{\mathbf{C}}^{-1} \tilde{\mathbf{H}}$  is block diagonal,  $\tilde{\mathbf{I}} \tilde{\mathbf{I}}^H$  is block diagonal

The EVD of a blockdiagonal matrix is also block diagonal, since

$$\begin{bmatrix} \tilde{\mathbf{A}} & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\mathbf{B}} \end{bmatrix} = \begin{bmatrix} \tilde{\mathbf{U}}_1 \tilde{\Lambda}_1 \tilde{\mathbf{U}}_1^H & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\mathbf{U}}_2 \tilde{\Lambda}_2 \tilde{\mathbf{U}}_2^H \end{bmatrix} = \begin{bmatrix} \tilde{\mathbf{U}}_1 & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\mathbf{U}}_2 \end{bmatrix} \begin{bmatrix} \tilde{\Lambda}_1 & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\Lambda}_2 \end{bmatrix} \begin{bmatrix} \tilde{\mathbf{U}}_1^H & \tilde{\mathbf{0}} \\ \tilde{\mathbf{0}} & \tilde{\mathbf{U}}_2^H \end{bmatrix} \quad A, B \text{ hermitian.}$$

Thus, the EVD of  $\tilde{\mathbf{H}}^H \tilde{\mathbf{C}}^{-1} \tilde{\mathbf{H}}$  and  $\tilde{\mathbf{I}} \tilde{\mathbf{I}}^H$  are also block diagonal

$\Rightarrow$  apply water-filling!

The  $\tilde{\mathbf{I}}$  can be found by  $\tilde{\mathbf{I}} = \tilde{\mathbf{U}} \tilde{\Psi}^{\frac{1}{2}}$ , where  $\tilde{\mathbf{U}}$  is the modal matrix of  $\tilde{\mathbf{H}} \tilde{\mathbf{C}}^{-1} \tilde{\mathbf{H}}^H$  ( $\tilde{\mathbf{U}}$  is also block diagonal),  $\tilde{\Psi}$  is the parameter found by ~~water-filling~~ waterfilling. The respective  $T_k$  can be found by identifying the corresponding position in the blockdiagonal matrix  $\tilde{\mathbf{I}}$ .

Q13)

- ① Optimal power allocation can achieve about twice the code rate of ~~at low and~~ equal power allocation at low and middle SNR.
- ② At high SNR, ~~the two schemes tend to give the same performance~~  
The difference between optimal power allocation and equal power allocation gets smaller, but optimal power allocation is still better than equal power allocation (about 0.3475 bits better ~~at~~ at 40 dB)