

Project 1. (Group Member: Jiangqing Li (alone))

Q1 - For k active stream

$$\gamma_k = \mu' - \frac{1}{\phi_k} > 0 \quad \gamma_{k+1} = 0 \quad \text{since} \quad \mu' - \frac{1}{\phi_{k+1}} < 0$$

$$\mu' = \frac{1}{k} (P_{tx} + \sum_{i=1}^k \frac{1}{\phi_i})$$

$$\frac{1}{\phi_k} < \frac{1}{k} (P_{tx} + \sum_{i=1}^k \frac{1}{\phi_i}) < \frac{1}{\phi_{k+1}}$$

$$\frac{k}{\phi_k} - \sum_{i=1}^k \frac{1}{\phi_i} < P_{tx} < \frac{k}{\phi_{k+1}} - \sum_{i=1}^k \frac{1}{\phi_i}$$

If $P_{tx} \geq \frac{k}{\phi_{k+1}} - \sum_{i=1}^k \frac{1}{\phi_i}$, then there can be $k+1$ active stream

$$\text{For } P_{tx} \rightarrow \infty \quad \gamma_i = \max(0, \frac{1}{k} P_{tx} + \frac{1}{k} \sum_{i=1}^k \frac{1}{\phi_i} - \frac{1}{\phi_i})$$

$$\lim_{P_{tx} \rightarrow \infty} \gamma_i = \frac{P_{tx}}{k}$$

Q2

$$P_{tx} = [0.1504, 2.3402, 26.2019]^T$$

Q3 For k active stream

$$\gamma_k = \frac{\bar{\mu}}{\sqrt{\phi_k}} - \frac{1}{\phi_k} > 0 \quad \gamma_{k+1} = 0, \text{ since } \frac{\bar{\mu}}{\sqrt{\phi_{k+1}}} - \frac{1}{\phi_{k+1}} < 0$$

$$\bar{\mu} = \frac{P_{tx} + \sum_{i=1}^k \frac{1}{\phi_i}}{\sum_{i=1}^k \frac{1}{\sqrt{\phi_i}}}$$

$$\frac{1}{\sqrt{\phi_k}} < \frac{P_{tx} + \sum_{i=1}^k \frac{1}{\phi_i}}{\sum_{i=1}^k \frac{1}{\sqrt{\phi_i}}} < \frac{1}{\sqrt{\phi_{k+1}}}$$

$$\frac{1}{\sqrt{\phi_k}} \sum_{i=1}^k \frac{1}{\sqrt{\phi_i}} - \sum_{i=1}^k \frac{1}{\phi_i} < P_{tx} < \frac{1}{\sqrt{\phi_{k+1}}} \sum_{i=1}^k \frac{1}{\sqrt{\phi_i}} - \sum_{i=1}^k \frac{1}{\phi_i}$$

If $P_{tx} \geq \frac{1}{\sqrt{\phi_{k+1}}} \sum_{i=1}^k \frac{1}{\sqrt{\phi_i}} - \sum_{i=1}^k \frac{1}{\phi_i}$, then there can be $k+1$ active stream.

$$P_{tx} \rightarrow \infty \quad \lim_{P_{tx} \rightarrow \infty} \bar{\mu} = \frac{P_{tx}}{\sum_{i=1}^k \frac{1}{\sqrt{\phi_i}}} \quad \lim_{P_{tx} \rightarrow \infty} \gamma_i = \frac{P_{tx}}{\sqrt{\phi_i} \sum_{j=1}^k \frac{1}{\sqrt{\phi_j}}} = \frac{\bar{\mu}}{\sqrt{\phi_i}}$$

Q4

$$P_{tx} = [0.0576, 0.5892, 4.2952]^T$$

Q5:

$$\begin{aligned}
 R_k &= \log_2 \det(I_N + H^H (C_n^{-1})^H Q) & H^H (C_n^{-1})^H H &= V \Phi V^H & Q &= V \bar{\Gamma} V^H \\
 &= \log_2 \det(I_N + V \Phi V^H V \bar{\Gamma} V^H) & \Phi, \bar{\Gamma} & \text{diagonal} \\
 &= \log_2 \det(I_N + \Phi \bar{\Gamma}) & \text{since } \gamma_1 \sim \gamma_k &= \frac{P_{Tx}}{K} & \gamma_{k+1} \sim \gamma_N &= 0 \\
 &= \sum_{i=1}^K \log_2 \left(1 + \phi_i \frac{P_{Tx}}{K} \right)
 \end{aligned}$$

Q6:

For uniform rate allocation

$$K=1 \quad P_{Tx} = 0.3009$$

$$K=2 \quad P_{Tx} = 4.4216$$

$$K=3 \quad P_{Tx} = 48.1417$$

For waterfilling

$$K=1 \quad P_{Tx} = 0.1504$$

$$K=2 \quad P_{Tx} = 2.3402$$

$$K=3 \quad P_{Tx} = 26.2019$$

• waterfilling solution switches earlier from K to $K+1$

• Yes, since when $P_{Tx} \rightarrow \infty$, waterfilling solution also allocates the transmission power uniformly, just like uniform rate allocation

Q7

$$\hat{s} = g H \underline{s} + g \underline{n}$$

$$\begin{aligned} \varepsilon(T, g) &= E[\|\underline{s} - \hat{s}\|_2^2] = E[\|\underline{s} - g H \underline{s} - g \underline{n}\|_2^2] = E[\|(\underline{I}_B - g H) \underline{s} - g \underline{n}\|_2^2] \\ &= E[(\underline{I}_B - g H) \underline{s} - g \underline{n}]^H (\underline{I}_B - g H) \underline{s} - g \underline{n}) \\ &= E[\text{tr}((\underline{I}_B - g H) \underline{s} - g \underline{n})^H (\underline{I}_B - g H) \underline{s} - g \underline{n})] \\ &= E[\text{tr}((\underline{I}_B - g H) \underline{s} - g \underline{n}) (\underline{I}_B - g H) \underline{s} - g \underline{n})^H] \\ &= \text{tr}(E[(\underline{I}_B - g H) \underline{s} - g \underline{n}) (\underline{I}_B - g H) \underline{s} - g \underline{n})^H]) \\ &= \text{tr}(E[(\underline{I}_B - g H) \underline{s} \underline{s}^H (\underline{I}_B - g^* H^H H^H) + |g|^2 \underline{n} \underline{n}^H]) \text{ since } E[\underline{s} \underline{n}^H] = 0 \\ &= \text{tr}((\underline{I}_B - g H) (\underline{I}_B - g^* H^H H^H) + |g|^2 \underline{C}_n) \\ &= \text{tr}((\underline{I}_B - g^* H^H H^H) (\underline{I}_B - g H) + |g|^2 \underline{C}_n) \\ &= \text{tr}(\underline{I}_B - g H \underline{I} - g^* \underline{I}^H H^H + |g|^2 \underline{I}^H H^H H \underline{I} + |g|^2 \underline{C}_n) \\ &= B - g \text{tr}(H \underline{I}) - g^* \text{tr}(\underline{I}^H H^H) + |g|^2 \text{tr}(\underline{I}^H H^H H \underline{I} + \underline{C}_n) \\ &= B - g \text{tr}(H \underline{I}) - g^* \text{tr}(\underline{I}^H H^H) + g g^* \text{tr}(\underline{I}^H H^H H \underline{I} + \underline{C}_n) \end{aligned}$$

Q8:

$$\frac{\partial \varepsilon(T, g)}{\partial g} = -\text{tr}(H \underline{I}) + g^* \text{tr}(\underline{I}^H H^H H \underline{I} + \underline{C}_n) \stackrel{!}{=} 0$$

$$g^* = \frac{\text{tr}(H \underline{I})}{\text{tr}(\underline{I}^H H^H H \underline{I} + \underline{C}_n)} \Rightarrow g = \frac{\text{tr}(\underline{I}^H H^H)}{\text{tr}(\underline{I}^H H^H H \underline{I} + \underline{C}_n)}$$

Since $\varepsilon(T, g)$ is convex in g , g is also a global minimum.

$$g = \frac{\text{tr}(\underline{I}^H H^H)}{\text{tr}(\underline{I}^H H^H H \underline{I} + \underline{C}_n)}$$

Q9:

$$E[\|\underline{x}\|_2^2] = E[\text{tr}(\underline{x} \underline{x}^H)] = E[\text{tr}(\underline{x} \underline{x}^H)] = \text{tr}(E[\underline{x} \underline{x}^H]) = \text{tr}(E[\underline{T} \underline{s} \underline{s}^H \underline{T}^H]) = \text{tr}(\underline{T} \underline{T}^H)$$

$$\text{tr}(\underline{T} \underline{T}^H) \leq P_{Tx}$$

$$L(T, \mu) = B - g \text{tr}(H \underline{I}) - g^* \text{tr}(\underline{I}^H H^H) + g g^* \text{tr}(\underline{I}^H H^H H \underline{I} + \underline{C}_n) + \mu (\text{tr}(\underline{T} \underline{T}^H) - P_{Tx})$$

KKT conditions:

$$\frac{\partial L(T, \mu)}{\partial T} = 0 \quad \mu (\text{tr}(\underline{T} \underline{T}^H) - P_{Tx}) = 0 \quad \mu \geq 0$$

$$\text{tr}(\underline{T} \underline{T}^H) - P_{Tx} \leq 0$$

Q10:

$$\frac{\partial \mathcal{L}(T, \mu)}{\partial T} = -g \mathbf{H}^T + |g|^2 (\mathbf{I}_N \mathbf{H}^H \mathbf{H})^T + \mu \cdot (\mathbf{T}^H)^T \stackrel{!}{=} 0$$

$$\mathbf{T}^H (\mu \mathbf{I}_N + |g|^2 \mathbf{H}^H \mathbf{H}) = g \mathbf{H} \quad \mathbf{T}^H = g \mathbf{H} (\mu \mathbf{I}_N + |g|^2 \mathbf{H}^H \mathbf{H})^{-1}$$

$$\mathbf{I} = g^* (\mu \mathbf{I}_N + |g|^2 \mathbf{H}^H \mathbf{H})^{-1} \mathbf{H}^H$$

Q11: $\mu (\text{tr}(\mathbf{T} \mathbf{T}^H) - P_{Tx}) = 0 \quad \mu > 0 \quad \text{tr}(\mathbf{T} \mathbf{T}^H) - P_{Tx} = 0$

$$-|g|^2 \text{tr}(\mathbf{C}_N) + \mu P_{Tx} = 0 \quad \mu = |g|^2 \frac{\text{tr}(\mathbf{C}_N)}{P_{Tx}}$$

$$\mathbf{I} = g^* (\mu \mathbf{I}_N + |g|^2 \mathbf{H}^H \mathbf{H})^{-1} \mathbf{H}^H$$

$$= g^* \left(|g|^2 \frac{\text{tr}(\mathbf{C}_N)}{P_{Tx}} + |g|^2 \mathbf{H}^H \mathbf{H} \right)^{-1} \mathbf{H}^H$$

$$= \frac{g^*}{g} \left(\frac{\text{tr}(\mathbf{C}_N)}{P_{Tx}} + \mathbf{H}^H \mathbf{H} \right)^{-1} \mathbf{H}^H$$

$$\mathbf{I} = \frac{1}{g} \left(\mathbf{H}^H \mathbf{H} + \frac{\text{tr}(\mathbf{C}_N)}{P_{Tx}} \mathbf{I} \right)^{-1} \mathbf{H}^H$$

$$\text{tr}(\mathbf{I} \mathbf{I}^H) = P_{Tx}$$

$$\text{tr} \left(\left(\mathbf{H}^H \mathbf{H} + \frac{\text{tr}(\mathbf{C}_N)}{P_{Tx}} \mathbf{I} \right)^{-1} \mathbf{H}^H \mathbf{H} \left(\mathbf{H}^H \mathbf{H} + \frac{\text{tr}(\mathbf{C}_N)}{P_{Tx}} \mathbf{I} \right)^{-1} \right) = |g|^2 P_{Tx}$$

$$|g|^2 = \frac{1}{P_{Tx}} \text{tr} \left(\left(\mathbf{H}^H \mathbf{H} + \frac{\text{tr}(\mathbf{C}_N)}{P_{Tx}} \mathbf{I}_N \right)^{-2} \mathbf{H}^H \mathbf{H} \right)$$

Q12:

- Slopes at 30 dB: waterfilling: 1.3142 bit/channel dB
MMSE : 1.3142 bit/channel dB
- About 1.8 dB
- For high SNR, uniform power allocation performs as waterfilling.
The MMSE solution with transmit filter performs worst. Its rate is smaller than the others

Q13:

- Slopes at 30 dB: waterfilling: -0.0100 / dB
-0.0057 / dB
- About 2.7 dB
- For high SNR, uniform power allocation performs as good as MMSE with transmit filter.

Although MMSE with transmit filter is MSE larger than MMSE, it performs the second well