

Random matrix theory: lecture 8Rate-diversity tradeoff in multi-antenna channels

Back to the third scenario: the quantity of interest is

$$P_{\text{out}}(R) = \min_{Q \geq 0: \text{Tr} Q \leq P} P(\log \det(I + H Q H^*) < R)$$

We again assume that h_{jk} are iid $\sim N_{\mathbb{C}}(0, 1)$, and would like to perform the analysis of P_{out} in the high SNR regime ($P \rightarrow \infty$). [ref: Zheng-Tse 03]

In the ergodic case, we have seen that

$$C = \max_{Q \geq 0: \text{Tr} Q \leq P} \mathbb{E}(\log \det(I + H Q H^*)) = \mathbb{E}(\log \det(I + \frac{P}{n} K K^*)) \\ = \mathbb{E}(\sum_{j=1}^n \log(1 + P \lambda_j)) \approx \min(m, n) \cdot \log P \quad \text{as } P \rightarrow \infty$$

(as shown by a simple heuristics)

Let us therefore choose a target rate $R = r \log P$, for some $0 \leq r \leq \min(m, n)$. We expect that

$$P_{\text{out}}(r \log P) \approx P^{-d(r)} \quad \text{for some } d(r) > 0.$$

\Rightarrow definition: the diversity order $d(r)$ associated to a

$$\text{rate } r \text{ is: } d(r) := \lim_{P \rightarrow \infty} - \frac{\log(P_{\text{out}}(r \log P))}{\log P}$$

[= upper bound on the diversity order of the error probability arising from a given coding scheme]

Our aim now is to compute $d(r)$ for

H $m \times n$ matrix with $iid \sim N_{\mathbb{C}}(0, 1)$ entries and $m \geq n$.

Notation: $f(P) \doteq g(P)$ if $\lim_{P \rightarrow \infty} \frac{\log f(P)}{\log P} = \lim_{P \rightarrow \infty} \frac{\log g(P)}{\log P}$ [without loss of generality]

[same "diversity order" for f & g]

Lemma

\times $P_{\text{out}}(r \log P) \doteq \mathbb{P}(\log \det(I + P H^* H) < r \log P)$

Proof

choose $Q = \frac{P}{n} I$

\bullet $A(P) := \mathbb{P}(\log \det(I + \frac{P}{n} H^* H) < r \log P) \geq P_{\text{out}}(r \log P)$

\times \bullet for any $Q \geq 0$ st. $\text{Tr } Q \leq P$, we have $Q \leq P I$;

\times since $A \mapsto \log \det A$ is increasing on the set of positive definite matrices, this implies

$\log \det(I + H Q H^*) \leq \log \det(I + P H^* H)$ for any Q ,

so $P_{\text{out}}(r \log P) \geq \mathbb{P}(\log \det(I + P H^* H) < r \log P) := B(P)$

\bullet finally, $A(P) \doteq B(P)$; indeed:

$B(P) \leq A(P) = \mathbb{P}(\log \det(I + \frac{P}{n} H^* H) < r \log P)$

$\doteq \mathbb{P}(\log \det(I + P H^* H) < r \log(nP))$

$\doteq \mathbb{P}(\log \det(I + P H^* H) < r \log P) = B(P)$

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Bottom line: the hard optimization problem is gone!

Therefore,

$$P_{\text{out}}(r \log P) = P \left(\sum_{j=1}^n \log(1 + P \lambda_j) < r \log P \right)$$

ev. of H^*H
↓

$$= \int_{D_\lambda(r)} p(\lambda_1, \dots, \lambda_n) d\lambda_1 \dots d\lambda_n$$

where $p(\lambda_1, \dots, \lambda_n) = C_n \cdot \exp \left(-\sum_{j=1}^n \lambda_j + (m-n) \sum_{j=1}^n \log \lambda_j \right)$
 $\cdot \prod_{j < k} (\lambda_k - \lambda_j)^2$ (see lecture 3)

and $D_\lambda(r) = \left\{ \underbrace{0 \leq \lambda_1 \leq \dots \leq \lambda_n}_{\text{eigenvalues ordered in ascending order}} : \sum_{j=1}^n \log(1 + P \lambda_j) < r \log P \right\}$

Let us now make the following change of variables:

$$\lambda_j = P^{-\alpha_j} = e^{-\alpha_j \log P} \quad \Rightarrow \quad d\lambda_j = -(\log P) e^{-\alpha_j \log P} d\alpha_j$$

$\alpha_j \in \mathbb{R}$
 $\alpha_j \geq 0$

so

$$P_{\text{out}}(r \log P) = \int_{D_\alpha(r)} q(\alpha_1, \dots, \alpha_n) d\alpha_1 \dots d\alpha_n$$

where $q(\alpha_1, \dots, \alpha_n) = C_n \exp \left(-\sum_{j=1}^n P^{-\alpha_j} - (m-n) \sum_{j=1}^n \alpha_j \log P \right)$
 $\cdot \prod_{j < k} (P^{-\alpha_k} - P^{-\alpha_j})^2 (\log P)^n \exp \left(-\sum_{j=1}^n \alpha_j \log P \right)$

and $D_\alpha(r) = \left\{ \alpha_1 \geq \dots \geq \alpha_n : \sum_{j=1}^n \log(1 + P^{1-\alpha_j}) < r \log P \right\}$
 $(\alpha_j \in \mathbb{R})$

At this point, let us make a couple of observations:

a) as $P \rightarrow \infty$, $\exp(-P^{-\alpha_j})$ $\begin{cases} \text{decays super-polynomially to 0} & \text{if } \alpha_j < 0 \\ \text{tends to 1} & \text{if } \alpha_j \geq 0 \end{cases}$

so we may restrict the integral to $\alpha_j \geq 0 \forall j$

b) as $P \rightarrow \infty$, $\log(1 + P^{1-\alpha_j}) \approx \begin{cases} (1-\alpha_j) \log P & \text{if } \alpha_j \leq 1 \\ 0 & \text{if } \alpha_j > 1 \end{cases}$

so $\log(1 + P^{1-\alpha_j}) \approx \underbrace{(1-\alpha_j)^+}_{\text{positive part}} \log P$

We may therefore replace the domain of integration $D_\alpha(r)$

by $\tilde{D}_\alpha(r) = \left\{ \alpha_1 \geq \dots \geq \alpha_n \geq 0 : \sum_{j=1}^n (1-\alpha_j)^+ < r \right\}$

So $P_{\text{out}}(r \log P) = \int_{\tilde{D}_\alpha(r)} \tilde{q}(\alpha_1, \dots, \alpha_n) d\alpha_1 \dots d\alpha_n$

where $\tilde{q}(\alpha_1, \dots, \alpha_n) = C_n (\log P)^n \cdot e^{-(m-n+1) \sum_j \alpha_j \log P}$
 $\cdot \prod_{j < k} (P^{-\alpha_k} - P^{-\alpha_j})^2$

Now, since $\alpha_1 \geq \dots \geq \alpha_n$, we have

$$\begin{aligned} \times \prod_{j < k} (P^{-\alpha_k} - P^{-\alpha_j})^2 &\approx \prod_{j < k} P^{-2\alpha_k} = \prod_k P^{-2(k-1)\alpha_k} \\ &= e^{-2 \sum_k (k-1) \alpha_k \log P} \end{aligned}$$

Note also that in terms of "diversity order", $(\log P)^n$ is

equivalent to a constant, since $\lim_{P \rightarrow \infty} \frac{\log((\log P)^n)}{\log P} = 0$
 (which is in turn equivalent to 1)

Finally,

$$P_{\text{out}}(r \log P) \stackrel{:= f(\alpha)}{=} \int_{\tilde{D}_\alpha(r)} e^{-\sum_j (m-n+2j-1) \alpha_j \log P} d\alpha_1 \dots d\alpha_n$$

$$= \int_{\tilde{D}_\alpha(r)} P^{-f(\alpha)} d\alpha_1 \dots d\alpha_n \stackrel{\text{Laplace integration method}}{=} P^{-\min_{\tilde{D}_\alpha(r)} f(\alpha)}$$

The final answer is therefore: $P_{\text{out}}(r \log P) = P^{-d(r)}$, where

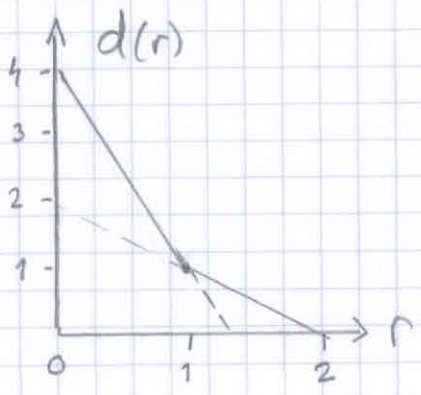
$$d(r) = \min_{\tilde{D}_\alpha(r)} f(\alpha) = \min_{\{\alpha_1 \geq \dots \geq \alpha_n \geq 0: \sum_{j=1}^n (1-\alpha_j)^+ < r\}} \sum_{j=1}^n (m-n+2j-1) \alpha_j$$

Ex: m=n=2

$$d(r) = \min_{\{\alpha_1 \geq \alpha_2 \geq 0: (1-\alpha_1)^+ + (1-\alpha_2)^+ < r\}} \alpha_1 + 3\alpha_2$$

0 < r < 1: $\alpha_1 = 1$ & $\alpha_2 = 1-r$ & $d(r) = 4-3r$
 i.e. $\lambda_1 \sim P^{-1}$ & $\lambda_2 \sim P^{r-1}$

1 < r < 2: $\alpha_1 = 2-r$ & $\alpha_2 = 0$ & $d(r) = 2-r$
 i.e. $\lambda_1 \sim P^{r-2}$ & $\lambda_2 \sim 1$



tradeoff:

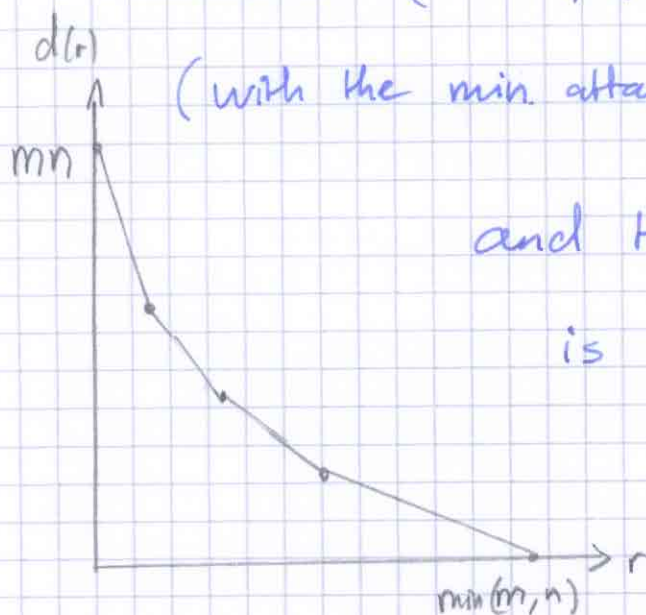
- { r ~ 2: full rate, low diversity
- { r ~ 0: low rate, full diversity

Picture in the general case

It turns out that for integer $r=k$, the solution of the above minimization problem is:

$$d(k) = (m-k)(n-k)$$

(with the min. attained at $\alpha_1 = \dots = \alpha_{n-k} = 1$, $\alpha_{n-k+1} = \dots = \alpha_n = 0$)



and the function in between

is piecewise linear (& convex)

Interpretation

With m receive antennas and n transmit antennas,

x a rate $k \log P$ can be "achieved" if $k \leq \min(m, n)$;

k antennas on each side contribute to this rate, while

the remaining $m-k$ and $n-k$ antennas provide diversity.

NB: for a target rate $R = k \log P$, the outage event is the event that the channel matrix H gives rise to less than k independent reliable scalar channels.