

Effect of Channel Correlation on the Performance of Diversity Receivers

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Introduction

- Fading: random fluctuations of the received signal power level in a wireless communication environment owing to
 - multipath propagation
 - relative movement between transmitter and receiver
- Diversity at the receiver: combats the effects of signal fading in mobile wireless channels
- Diversity schemes:
 - Space diversity: use multiple antennas to receive same signal

- Frequency diversity: transmit the same signal over different frequencies
- Time diversity: Same signal is transmitted at multiple instants
- Path diversity: used in CDMA systems; resolve multipaths via code correlation and combine them in a rake receiver
- Different kinds of fading statistics of the amplitude of the signal of each diversity branch:
 - Rayleigh fading (diffuse component only, no specular or LOS component)
 - Rician fading (diffuse component and at most one specular component per branch)
 - Nakagami fading (superimposition of several Rayleigh faded diffuse components)

- Diversity combining techniques:
 - Maximal-ratio combining (MRC) (in presence of interference, use optimum combining (OC))
 - Predetection equal-gain combining (EGC)
 - Selection combining (SC)
 - Hybrid selection/maximal-ratio combining (H-S/MRC)
 - Postdetection equal-gain combining (PEGC)
- MRC, EGC, H-S/MRC need coherent reception
- SC can be used with coherent, noncoherent, or differentially coherent reception
- PEGC is used with noncoherent or differentially coherent reception

BNFSK and BDPSK with PEGC over Correlated Rician Channels

- Although independent fading between diversity branches is desirable, the fading can be correlated in many practical scenarios
- Consider a diversity reception system over a flat Rician fading channel with L correlated branches
- Receiver employs symbol-by-symbol detection
- Signal received over the k th diversity branch in a symbol interval of duration T_s :

$$r_k(t) = \text{Re} \left\{ \left[\alpha_k e^{-j\phi_k} s(t) + n_k(t) \right] e^{j2\pi f_c t} \right\},$$
$$k = 1, \dots, L, \quad 0 \leq t < T_s$$

- $s(t)$: complex baseband information-bearing signal with average symbol energy $2E_s$
- α_k, ϕ_k : random magnitude and random phase of k th diversity branch gain
- f_c : carrier frequency
- $n_k(t)$: additive noise (zero-mean complex white Gaussian random process with two-sided power spectral density $2N_0$)
- $\{n_k(t)\}$ are independent but $\{\alpha_k e^{-j\phi_k}\}$ are correlated
- Each $n_k(t)$ is independent of the gains $\{\alpha_k e^{-j\phi_k}\}$
- Complex channel gain vector:

$$\underline{g} \triangleq [\alpha_1 e^{-j\phi_1}, \dots, \alpha_L e^{-j\phi_L}]^T = \underline{X}_c + j\underline{X}_s$$

- \underline{X}_c : in-phase component
 \underline{X}_s : quadrature component
- \underline{X}_c and \underline{X}_s are real Gaussian vectors with mean vectors $\underline{\mu}_c$ and $\underline{\mu}_s$, covariance matrices \underline{K}_{cc} and \underline{K}_{ss} , cross-covariance matrix \underline{K}_{cs}
- \underline{K}_{cc} and \underline{K}_{ss} have the same diagonal elements, and all diagonal elements of \underline{K}_{cs} are zero
- $\mu_{c_k} + j\mu_{s_k}$: specular component of g_k

- Now

$$\underline{X} \triangleq \begin{bmatrix} \underline{X}_c \\ \underline{X}_s \end{bmatrix} \sim \mathcal{N}(\underline{\mu}, \underline{K}),$$

where

$$\underline{K} \triangleq \begin{bmatrix} \underline{K}_{cc} & \underline{K}_{cs} \\ \underline{K}_{cs}^T & \underline{K}_{ss} \end{bmatrix}, \quad \underline{\mu} \triangleq \begin{bmatrix} \underline{\mu}_c \\ \underline{\mu}_s \end{bmatrix}$$

- Let $\underline{K} = \underline{E} \underline{\Lambda} \underline{E}^T$, where $\underline{\Lambda} = \text{diag}(\lambda_1, \dots, \lambda_{2L})$
- $\underline{\Lambda}$: diagonal matrix of eigenvalues of \underline{K} ,
 \underline{E} : orthogonal matrix of eigenvectors of \underline{K}
- $\underline{\eta} = [\eta_1, \dots, \eta_{2L}]^T = \underline{E}^T \underline{\mu}$
- For BNFSK, signal for symbol i :

$$s(t) = \sqrt{\frac{2E_s}{T_s}} e^{j2\pi(\Delta f_i)t}, \quad i = 0, 1$$

- $r_k(t)$ is passed through two narrowband bandpass filters (BPFs)
- u_{ki} : square of output envelope for BPF centered at $f_c + \Delta f_i$
- Decision variable : $D = \sum_{k=1}^L u_{k1} - \sum_{k=1}^L u_{k0}$
- When symbol 1 is transmitted:

$$u_{k1} = |2E_s g_k + N_{k1}|^2, \quad u_{k0} = |N_{k0}|^2$$
- N_{k1} and N_{k0} are i.i.d. $\mathcal{CN}(0, 4E_s N_0)$ random variables
- Similar approach for BDPSK

- $\gamma_k = \frac{E_s \alpha_k^2}{N_0}$: instantaneous SNR at branch k
- $\gamma_{tot} = \sum_{k=1}^L \gamma_k$: instantaneous SNR at
combiner output
- By using results on quadratic forms in normal variables, we find the c.f. of γ_{tot} and then the c.f. of D
- SEP = Pr[$D < 0$ when symbol 1 is transmitted]
- The SEP is obtained directly from the c.f. of D using the inversion theorem (Gil-Pelaez lemma)

- Define

$$a_k \triangleq 2 + 4h\lambda_k \frac{E_s}{N_0}, \quad b_k \triangleq 2h\eta_k^2 \frac{E_s}{N_0},$$

where

$$h = \begin{cases} \frac{1}{2} & \text{for BNFSK,} \\ 1 & \text{for BDPSK} \end{cases}$$

- SEP:

$$P_e = \left[\prod_{k=1}^{2L} \frac{\exp\left\{-\frac{b_k}{a_k}\right\}}{a_k^{\frac{1}{2}}}\right] \times \sum_{\substack{(l_1, \dots, l_{L-1}) \\ 0 \leq l_1, \dots, l_{L-1} \leq L-1 \\ l_1 + 2l_2 + \dots + (L-1)l_{L-1} = L-1}} \prod_{m=1}^{L-1} \frac{1}{l_m!} \left[\frac{1}{m} + \frac{1}{2m} \sum_{k=1}^{2L} \left(1 - \frac{1}{a_k}\right)^m + \sum_{k=1}^{2L} \frac{b_k}{a_k^2} \left(1 - \frac{1}{a_k}\right)^{m-1} \right]^{l_m}$$

- Numerical results for exponential correlation type of power balanced diversity:

$$\underline{K}_{cs} = \underline{0}, \quad (\underline{K}_{cc})_{ij} = (\underline{K}_{ss})_{ij} = \rho^{|i-j|} \sigma_Z^2$$

- ρ : correlation coefficient ($0 \leq \rho < 1$)
- Phase difference of ξ between specular components of k th and $(k + 1)$ th branches:

$$(\underline{\mu}_c + j\underline{\mu}_s)_k = \mu_0 e^{j(k-1)\xi}$$

- $\frac{E_s}{N_0}(\mu_0^2 + 2\sigma_Z^2)$: branch SNR
- $K = \frac{\mu_0^2}{2\sigma_Z^2}$: Rician factor

- For all ξ satisfying $0 \leq |\xi| < \xi_{max}$ (ξ_{max} depends on ρ , L , and branch SNR):
 - SEP versus K reaches a maximum at some $K = K_{max}$
 - K_{max} decreases as $|\xi|$ goes from 0 to ξ_{max}
 - For $|\xi| \geq \xi_{max}$, SEP attains a maximum at $K = 0$, which corresponds to Rayleigh fading
- In the case of correlated branches, for a certain range of values of ξ , the performance in Rayleigh fading is better than that in Rician fading over a range $0 < K < K_m$
- However, in the case of independent branches the performance in Rayleigh fading is worse than that in Rician fading

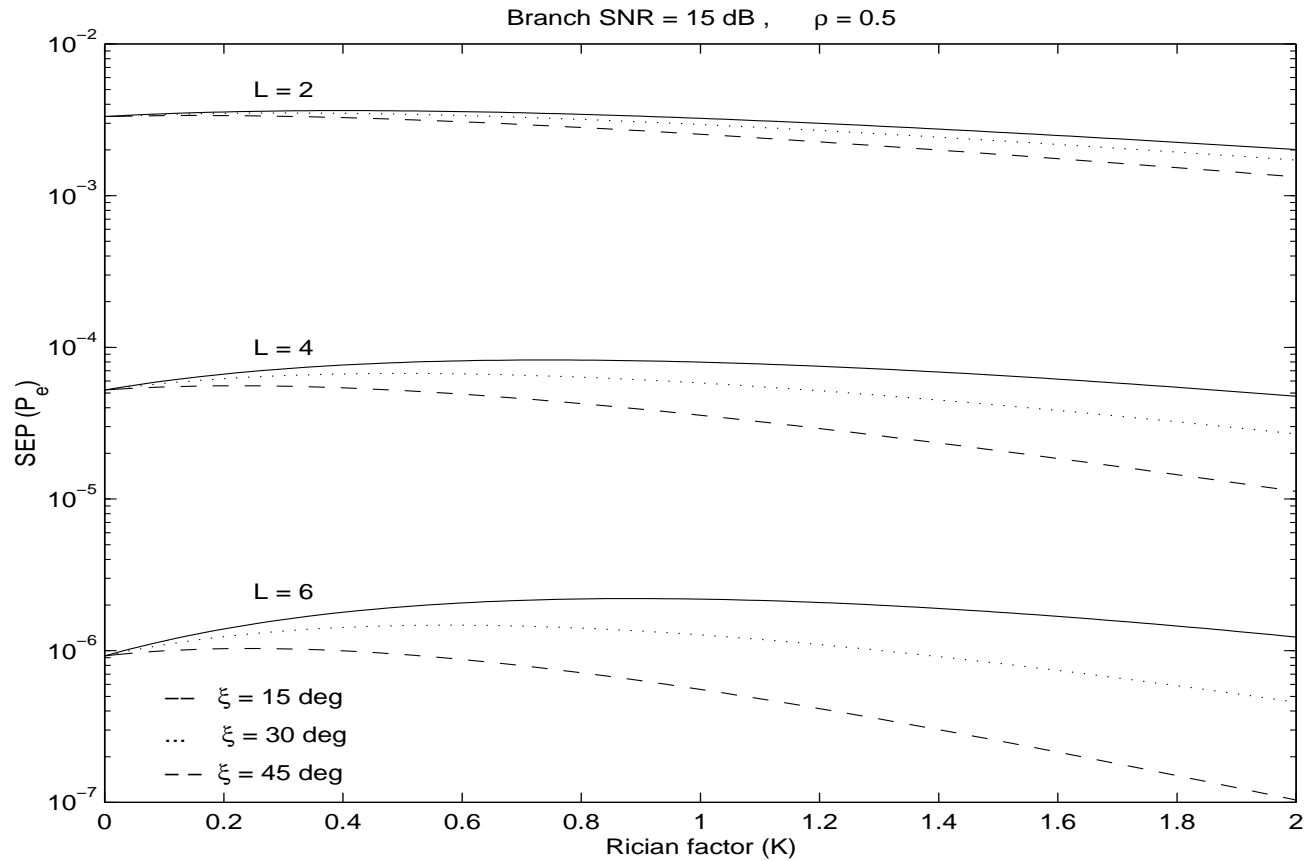


Figure 1: SEP versus Rician factor when branch SNR = 15 dB, $\rho = 0.5$ ($\xi_{max} \approx 60^\circ$) for different values of L and ξ in case of BNFSK using exponential correlation type power balanced diversity

H-S/MRC in Correlated Nakagami Fading

- Nakagami fading model with evenly correlated diversity branches having the same m -parameter and same average SNR Γ
- Select a subset L of N available branches with highest instantaneous SNR
- From the joint c.f. of the instantaneous branch SNRs, an expression for the c.f. of the combiner output SNR is obtained
- This c.f. is used to obtain the SEP of coherent detection several M -ary modulation schemes
- Properties of order statistics are applied

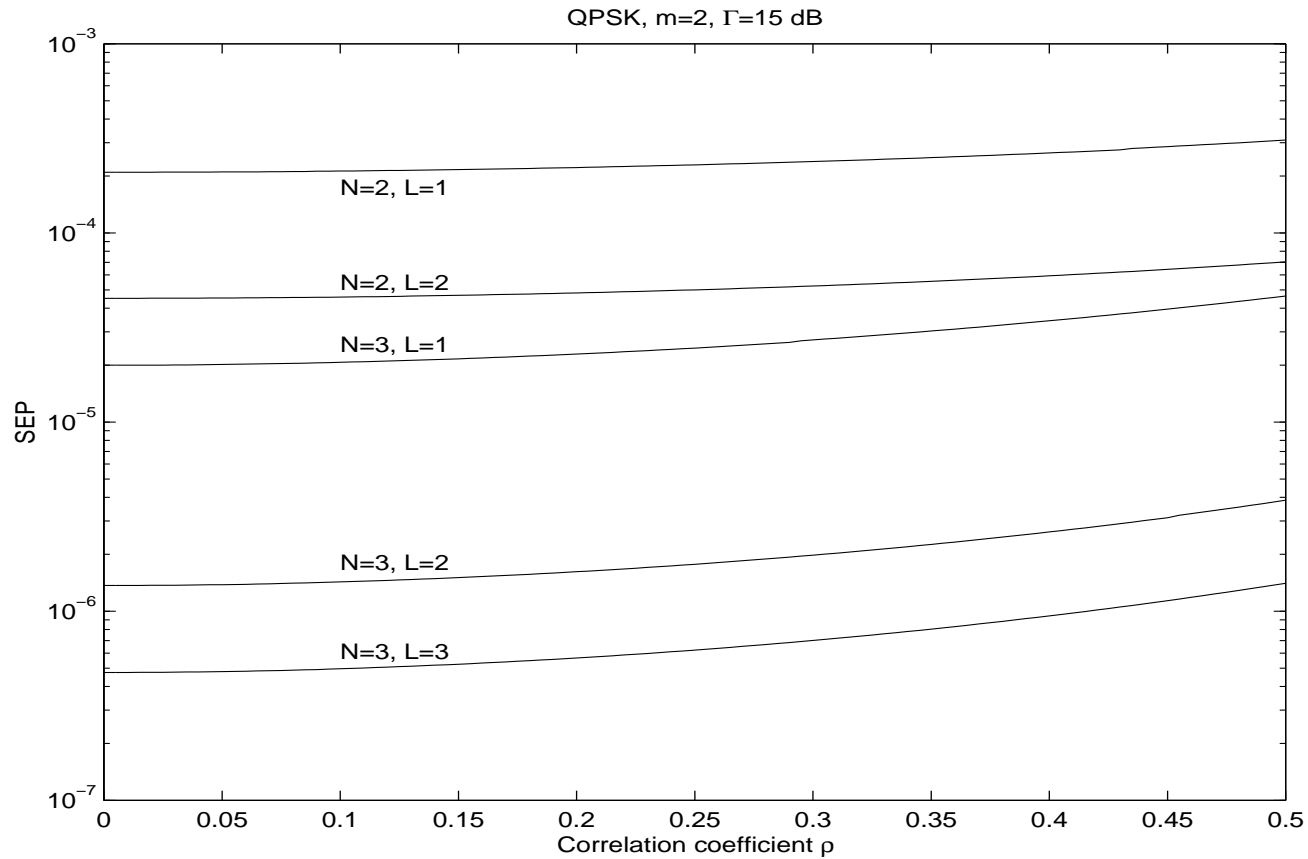


Figure 2: SEP as a function of correlation coefficient ρ for correlated Nakagami fading having parameter $m = 2$, average SNR per branch $\Gamma = 15$ dB, $1 \leq L \leq N$, $N = 2, 3$

Effect of Channel Correlation

- Correlation degrades the performance
- In case of PEGC, we do not estimate the channel; and we cannot do much to combat the effect of correlation
- In case of H-S/MRC, we have assumed that perfect estimates of $\{\alpha_k e^{-j\phi_k}\}$ are available to the receiver
- In practice, there will be some estimation error, and this will further degrade the performance
- Schemes which need channel estimation can be modified to yield better performance in presence of channel correlation

Optimized Diversity Combining

- In the presence of CCI, OC is known to give the best error performance since it maximizes the instantaneous SINR of the combiner output
- This is based on the assumption that a perfect estimate of the channel is available
- With imperfect channel estimation, the receiver structure needs to be modified to combat the effect of channel correlation
- This results in what we call optimized diversity combining (ODC)
- Consider a pilot symbol based ML channel estimation method

- Correlated flat Rayleigh fading channel in the presence of CCI and additive noise
- The decision rule, which is optimum in the ML sense, is derived using concepts of Gaussian and Wishart statistics
- A system using receive diversity with L receive antennas, and subject to N cochannel interferers
- We use K pilot symbols to estimate the channel
- Complex baseband sampled received signal vector for k th pilot transmission:

$$\mathbf{r}_{p_k} = \mathbf{c}s_{p_k} + \sum_{i=1}^N \mathbf{b}_i q_{p_i,k} + \mathbf{n}_{p_k}, \quad k = 1, \dots, K$$

- \mathbf{c} : propagation vector of desired user
 s_{p_k} : pilot symbol
 $\{\mathbf{b}_i\}$: propagation vectors of interferers
 $\{q_{p_{i,k}}\}$: their signal samples
 \mathbf{n}_{p_k} : AWGN vector
- $\{q_{p_{i,k}}\}_{i,k}$ are i.i.d. $\mathcal{CN}(0, 1)$ random variables
- $\mathbf{n}_{p_1}, \dots, \mathbf{n}_{p_K}$ are i.i.d. $\mathcal{CN}(\mathbf{0}, \sigma^2 \mathbf{I}_L)$ random vectors
- $\mathbf{c} \sim \mathcal{CN}(\mathbf{0}, \mathbf{K}_c)$
 $\mathbf{b}_1, \dots, \mathbf{b}_N$ are i.i.d. $\mathcal{CN}(\mathbf{0}, \mathbf{K}_b)$
- $\mathbf{c}, \mathbf{b}_1, \dots, \mathbf{b}_N, \{q_{p_{i,k}}\}_{i,k}$, and $\mathbf{n}_{p_1}, \dots, \mathbf{n}_{p_K}$ are independent of each other

- Receiver has knowledge of \mathbf{K}_c , \mathbf{K}_b , and σ^2 which characterize the channel statistics
- Given the channel parameters \mathbf{c} , $\mathbf{b}_1, \dots, \mathbf{b}_N$, the received signal vectors $\mathbf{r}_{p_1}, \dots, \mathbf{r}_{p_K}$ are independent, \mathbf{r}_{p_k} having a

$$\mathcal{CN} \left(\mathbf{c} s_{p_k}, \sum_{i=1}^N \mathbf{b}_i \mathbf{b}_i^H + \sigma^2 \mathbf{I}_L \right)$$

distribution

- Interference correlation matrix: $\mathbf{A} = \sum_{i=1}^N \mathbf{b}_i \mathbf{b}_i^H$
- Pilot symbol vector:

$$\mathbf{s}_p = \begin{bmatrix} s_{p_1} \\ \vdots \\ s_{p_K} \end{bmatrix}$$

- Received pilot vector:

$$\mathbf{r}_p = \begin{bmatrix} \mathbf{r}_{p_1} \\ \vdots \\ \mathbf{r}_{p_K} \end{bmatrix}$$

- The distribution of \mathbf{r}_p , conditioned on the channel parameters \mathbf{c} , \mathbf{A} , is

$$\mathcal{CN}(\mathbf{s}_p \otimes \mathbf{c}, \mathbf{I}_K \otimes (\mathbf{A} + \sigma^2 \mathbf{I}_L))$$

- \mathbf{A} is complex Wishart distributed when $N \geq L$, and complex pseudo-Wishart distributed when $N < L$
- Denote this distribution as $\mathcal{CW}_L(N, \mathbf{K}_b)$

- $\mathcal{M}_+(L)$: set of all $L \times L$ complex positive definite matrices
- $\mathcal{M}_+(L, N)$: set of all $L \times L$ complex Hermitian rank- N matrices each with positive definite $N \times N$ principal submatrices
- $\mathbf{A}_{[N]}$: first principal $N \times N$ submatrix of \mathbf{A}
- P.d.f. of \mathbf{A} :

$N \geq L$:

$$f(\mathbf{A}) = \frac{\{\det(\mathbf{A})\}^{N-L} \exp\{-\text{tr}(\mathbf{K}_b^{-1}\mathbf{A})\}}{\pi^{\frac{L(L-1)}{2}} \left\{ \prod_{i=1}^L (N-i)! \right\} \{\det(\mathbf{K}_b)\}^N}, \quad \mathbf{A} \in \mathcal{M}_+(L)$$

$N < L$:

$$f(\mathbf{A}) = \frac{\{\det(\mathbf{A}_{[N]})\}^{-(L-N)} \exp\{-\text{tr}(\mathbf{K}_b^{-1}\mathbf{A})\}}{\pi^{\frac{N(2L-N-1)}{2}} \left\{ \prod_{i=1}^N (N-i)! \right\} \{\det(\mathbf{K}_b)\}^N}, \quad \mathbf{A} \in \mathcal{M}_+(L, N)$$

- ML estimate of (\mathbf{c}, \mathbf{A}) :

$$(\hat{\mathbf{c}}, \hat{\mathbf{A}}) = \arg \left\{ \max_{\mathbf{c}, \mathbf{A}} \ln f(\mathbf{r}_p | \mathbf{c}, \mathbf{A}) \right\}$$

- Maximization of the log likelihood function with respect to \mathbf{c} , \mathbf{A} results in

$$\hat{\mathbf{c}} = \frac{1}{\left(\sum_{k=1}^K |s_{p_k}|^2 \right)} \sum_{k=1}^K s_{p_k}^* \mathbf{r}_{p_k},$$

$$\hat{\mathbf{A}} = \frac{1}{K} \sum_{k=1}^K (\mathbf{r}_{p_k} - \hat{\mathbf{c}} s_{p_k}) (\mathbf{r}_{p_k} - \hat{\mathbf{c}} s_{p_k})^H - \sigma^2 \mathbf{I}_L$$

- It can be shown that $\hat{\mathbf{c}}$ and $\hat{\mathbf{A}}$, conditioned on \mathbf{c} , \mathbf{A} , are independent, and that

$$\hat{\mathbf{c}}|_{\mathbf{c}, \mathbf{A}} \sim \mathcal{CN}\left(\mathbf{c}, \frac{1}{(\mathbf{s}_p^H \mathbf{s}_p)} (\mathbf{A} + \sigma^2 \mathbf{I}_L)\right),$$

$$(\hat{\mathbf{A}} + \sigma^2 \mathbf{I}_L)|_{\mathbf{c}, \mathbf{A}} \sim \mathcal{CW}_L\left(K - 1, \frac{1}{K} (\mathbf{A} + \sigma^2 \mathbf{I}_L)\right)$$

- Consider the reception of a transmitted data symbol s over the same channel after the reception of the K pilot symbols s_{p_1}, \dots, s_{p_K}
- Complex baseband sampled received signal vector for data transmission:

$$\mathbf{r} = \mathbf{c}s + \sum_{i=1}^N \mathbf{b}_i q_i + \mathbf{n}$$

- Interfering signal samples $\{q_i\}_i$ are i.i.d. $\mathcal{CN}(0, 1)$
- Additive noise vector \mathbf{n} has a $\mathcal{CN}(\mathbf{0}, \sigma^2 \mathbf{I}_L)$ distribution
- \mathbf{c} , $\mathbf{b}_1, \dots, \mathbf{b}_N$, $\{q_i\}_i$, and \mathbf{n} are independent of each other, as well as of the interfering signal samples $\{q_{p_i,k}\}_{i,k}$, and the noise vectors $\{\mathbf{n}_{p_i}\}$ corresponding to pilot symbol reception
- s belongs to an M -ary signal constellation, and takes one of M complex values S_1, \dots, S_M

- When perfect estimates of \mathbf{c} and \mathbf{A} are available, the optimum decision rule in the ML sense is given by

$$\hat{s} = \arg \left\{ \max_{s \in \{S_1, \dots, S_M\}} f(\mathbf{r} | \mathbf{c}, \mathbf{A}, s) \right\}$$

- This results in

$$\hat{s} = \arg \left\{ \max_{s \in \{S_1, \dots, S_M\}} \left[\operatorname{Re} \{s^* \mathbf{w}_{oc}^H \mathbf{r}\} - \frac{|s|^2}{2} \mathbf{c}^H (\mathbf{A} + \sigma^2 \mathbf{I}_L)^{-1} \mathbf{c} \right] \right\}$$

where $\mathbf{w}_{oc} = (\mathbf{A} + \sigma^2 \mathbf{I}_L)^{-1} \mathbf{c}$ is the OC weight vector

- On the other hand, when we have imperfect estimates $\hat{\mathbf{c}}$ of \mathbf{c} and $\hat{\mathbf{A}}$ of \mathbf{A} , the optimum decision rule in the ML sense is

$$\hat{s} = \arg \left\{ \max_{s \in \{S_1, \dots, S_M\}} f(\mathbf{r} | \hat{\mathbf{c}}, \hat{\mathbf{A}}, s) \right\}$$

- We can express the conditional p.d.f. of $\mathbf{r}, \hat{\mathbf{c}}$ as

$$\begin{aligned} & f(\mathbf{r}, \hat{\mathbf{c}} | \mathbf{A}, s) \\ &= \frac{1}{\pi^{2L} \det(|s|^2 \mathbf{K}_c + \mathbf{A} + \sigma^2 \mathbf{I}_L) \det(\mathbf{M}_{22}^{-1}(\mathbf{A}, s))} \\ & \quad \times \exp \left\{ - \left[\mathbf{r}^H \mathbf{M}_{11}(\mathbf{A}, s) \mathbf{r} + \hat{\mathbf{c}}^H \mathbf{M}_{22}(\mathbf{A}, s) \hat{\mathbf{c}} + 2 \operatorname{Re} \left(\hat{\mathbf{c}}^H \mathbf{M}_{12}^H(\mathbf{A}, s) \mathbf{r} \right) \right] \right\} \end{aligned}$$

- $\{M_{ij}(\mathbf{A}, s)\}$ are the blocks of the inverse of the covariance matrix of $\mathbf{r}, \hat{\mathbf{c}}$, and are given by

$$\mathbf{M}_{11}(\mathbf{A}, s) = \left(|s|^2 \mathbf{K}_c + \mathbf{A} + \sigma^2 \mathbf{I}_L - |s|^2 \mathbf{K}_c \left(\mathbf{K}_c + \frac{1}{(\mathbf{s}_p^H \mathbf{s}_p)} (\mathbf{A} + \sigma^2 \mathbf{I}_L) \right)^{-1} \mathbf{K}_c \right)^{-1},$$

$$\mathbf{M}_{22}(\mathbf{A}, s) = \left(\mathbf{K}_c + \frac{1}{(\mathbf{s}_p^H \mathbf{s}_p)} (\mathbf{A} + \sigma^2 \mathbf{I}_L) - |s|^2 \mathbf{K}_c (|s|^2 \mathbf{K}_c + \mathbf{A} + \sigma^2 \mathbf{I}_L)^{-1} \mathbf{K}_c \right)^{-1},$$

$$\mathbf{M}_{12}(\mathbf{A}, s) = -s \mathbf{M}_{11}(\mathbf{A}, s) \mathbf{K}_c \left(\mathbf{K}_c + \frac{1}{(\mathbf{s}_p^H \mathbf{s}_p)} (\mathbf{A} + \sigma^2 \mathbf{I}_L) \right)^{-1}$$

- The decision rule can be written as

$$\hat{s} = \arg \left\{ \max_{s \in \{S_1, \dots, S_M\}} g(\mathbf{r}, \hat{\mathbf{c}}, \hat{\mathbf{A}} | s) \right\},$$

where the sufficient statistic

$$g(\mathbf{r}, \hat{\mathbf{c}}, \hat{\mathbf{A}} | s) = \int_{\mathbf{A}} f(\mathbf{r}, \hat{\mathbf{c}} | \mathbf{A}, s) f(\hat{\mathbf{A}} | \mathbf{A}) f(\mathbf{A}) d\mathbf{A}$$

- Numerical results for MPSK with $|s| = 1$,

$$(\mathbf{K}_c)_{k,l} = \begin{cases} P_S \rho^{|k-l|} & \text{if } k \neq l, \\ P_S & \text{if } k = l, \end{cases}$$

P_S being the signal power, and

$$(\mathbf{K}_b)_{k,l} = \begin{cases} P_I \rho^{|k-l|} & \text{if } k \neq l, \\ P_I & \text{if } k = l \end{cases}$$

P_I being the interference power

- Average SNR per branch: P_S/σ^2
Average SIR per branch per interferer: P_S/P_I
Pilot-to-noise ratio (PNR): $\mathbf{s}_p^H \mathbf{s}_p/\sigma^2$
- Comparison of ODC with OCI (OC using imperfect channel estimates), OC, and MRC

- In case of ODC, the sufficient statistic has been computed by Monte Carlo simulation
- Performance of ODC is in between that of OCI and OC
- At an SEP of 7×10^{-3} , ODC has an advantage of about 3 dB over OCI and a disadvantage of only about 1 dB compared to OC when $N = 2$
- OCI is about 2 times as complex as OC, while ODC is about 100 times as complex as OCI
- ODC can perform significantly better than OCI with appropriate choice of system parameters

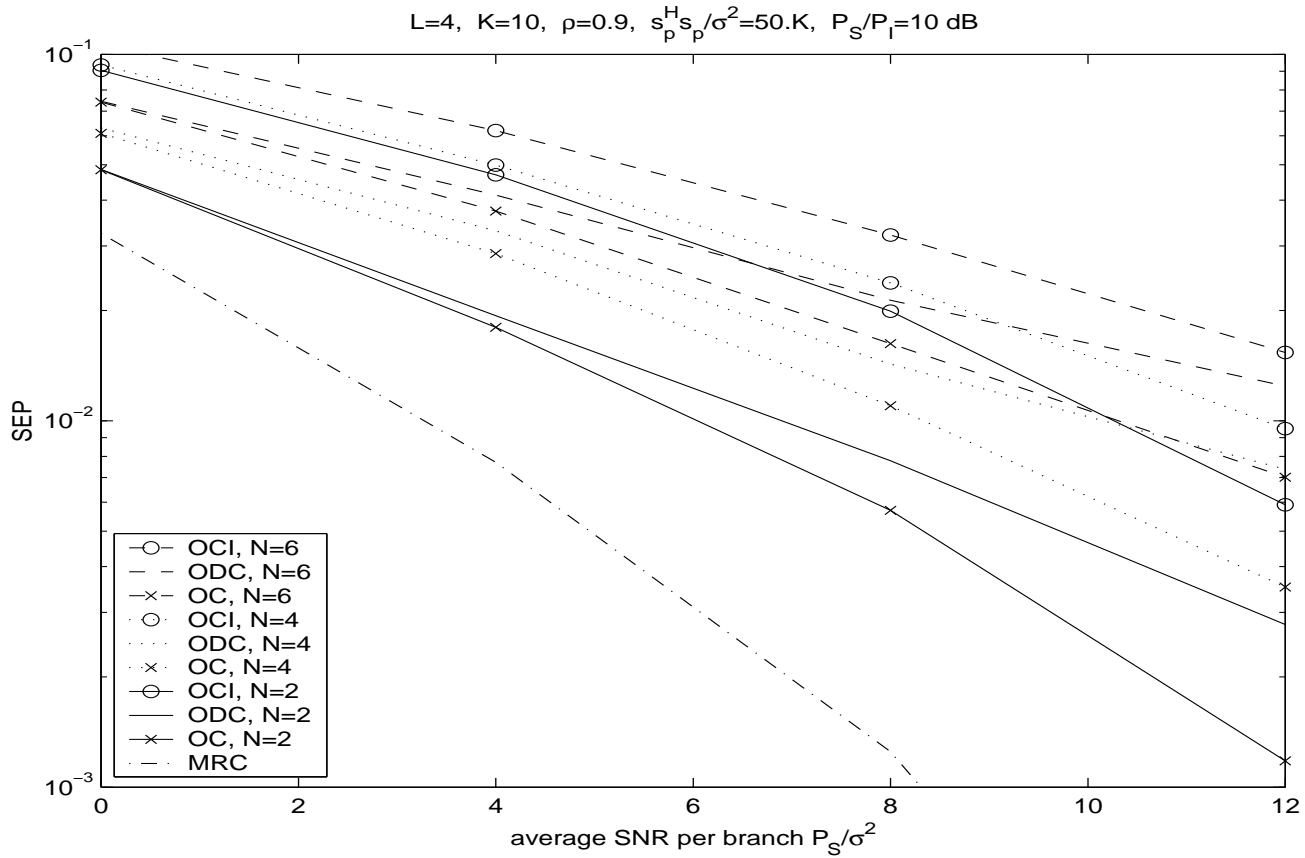


Figure 3: Comparison of error performance of ODC, OCI, and OC for BPSK with number of receive antennas $L = 4$, number of interferers $N = 2, 4, 6$, number of pilot symbols $K = 10$, exponentially correlated fading having $\rho = 0.9$, PNR $s_p^H s_p / \sigma^2 = 50 \cdot K$, and average SIR per branch per interferer $P_S/P_I = 10 \text{ dB}$

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