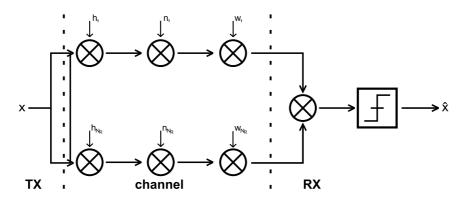
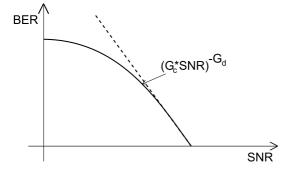
# 1 SIMO Systems

#### Remarks

- $\bullet$  In SIMO Systems only <u>coding</u> and <u>diversity</u> <u>gains</u> can be exploited (no multiplexing gains)
- To realize these gains diversity combining has to be performed
- Diversity combining schemes vary in complexity and performance
- There are many diversity combining schemes. Here we consider:
  - Maximal ratio combining (MRC)
  - Equal gain combining (EGC)
  - Selection combining (SC)
- Diversity combining problem



- how to choose combining weights  $w_n$ ?
- what performance (e.g. error rate, outage probability) is achieved?
- what diversity and coding/combining gain is achieved?



- $G_c$ : Coding gain
- $G_d$ : Diversity gain

### 1.1 Preliminaries

Consider an equivalent system:

$$y=hx+n;$$
 
$$\mathcal{E}\{|x^2|\}=\epsilon_s; \qquad \qquad \mathcal{E}\{|n^2|\}=\sigma_n^2; \qquad \qquad \mathcal{E}\{|h|^2\}=1$$

- Instantaneous SNR:  $\gamma_t = \frac{\epsilon_s}{\sigma_n^2} \times |h|^2$
- Average SNR:  $\bar{\gamma}_t = \mathcal{E}\{\gamma_t\} = \frac{\epsilon_s}{\sigma_n^2}$

#### Bit and Symbol Error Rate

• The Bit and Symbol Error Rate of many modulation schemes can be expressed for given  $\gamma_t$  as:

$$P_e(\gamma_t) = aQ\{\sqrt{b\gamma_t}\}$$

where:

- $Q(x) = \frac{1}{\sqrt{2\pi}} \times \int_x^\infty e^{-\frac{t^2}{2}} dt$
- $P_e(\gamma_t)$  may be exact result or approximation
- BPSK: exact with a = 1, b = 2
- M-ary QAM: tight approximation with  $a = 4\left(1 \frac{1}{\sqrt{M}}\right), b = \frac{3}{M-1}$

 $\left(Einschub: Gray - Code: BER = \frac{1}{\log_2 M} \times SER\right)$ 

• Alternative representation of Q - function:

$$Q(x) = \frac{1}{\pi} \int_0^{\frac{\pi}{2}} e^{-\frac{x^2}{2\sin^2\theta}} \ d\theta$$

- → Integral limits are fixed and do not depend on integration variables!
- Average error probability

$$P_e = \mathcal{E}\{P_e(\gamma_t)\} = \int_0^\infty aQ(\sqrt{bx})p_{\gamma_t}(x) dx$$

- Integral may be difficult to solve analytically
- Integral has infinite support  $\rightarrow$  numerical evaluation difficult
- Using alternative representation of Q-function we get:

$$P_{e} = \int_{0}^{\infty} \frac{a}{\pi} \int_{0}^{\frac{\pi}{2}} e^{-\frac{bx}{2sin^{2}\theta}} p_{\gamma_{t}}(x) d\theta dx$$

$$= \frac{a}{\pi} \int_{0}^{\frac{\pi}{2}} \int_{0}^{\infty} p_{\gamma_{t}}(x) e^{-\frac{b}{2sin^{2}\theta}} dx d\theta \qquad = \frac{a}{\pi} \int_{0}^{\frac{\pi}{2}} M_{\gamma_{t}}(\frac{b}{2sin^{2}\theta}) d\theta$$

where:

- $-M_{\gamma_t}(s) = \int_0^\infty p_{\gamma_t}(x)e^{-sx} dx$  is the Laplace transform of  $p_{\gamma_t}(s)$
- $-M_{\gamma_t}(-s)$  is the so called Moment Generation Function (MGF) of  $p_{\gamma_t}$
- Here, we will also refer to  $M_{\gamma_t}(s)$  as MGF
- $M_{\gamma_t}(s)$  is sometimes easier to obtain than  $p_{\gamma_t}$
- The above integral can be easily evaluated numerically because of the finite integral limits

### Outage probability

• The outage probability is the probability that the channel cannot support a certain rate, R, i.e. (where  $\gamma_T$  is the threshold SNR):

$$C = \log_2(1 + \gamma_t) < R \quad \leftrightarrow \quad \gamma_t < 2^R - 1 \triangleq \gamma_T$$

Thus, the outage probability is given by:

$$P_{out} = P_0 \gamma_t < \gamma - T = \int_0^{\gamma_T} p_{\gamma_t}(x) \ dx$$

• Using the inverse Laplace Transform

$$p_{\gamma_t}(x) = \frac{1}{2\pi j} \int_{c-j\omega}^{c+j\omega} M_{\gamma_t}(s) e^{sx} dx$$

where c > 0 is a small constant that lies in the region of convergence of the integral, we obtain:



- 1.

$$P_{out} = \frac{1}{2\pi j} \int_{c-j\omega}^{c+j\omega} M_{\gamma_t}(s) \int_0^{\gamma_T} e^{sx} dx ds = \frac{1}{2\pi j} \int_{c-j\omega}^{c+j\omega} M_{\gamma_t}(s) e^{\gamma_T s} \frac{ds}{s}$$

(lower integral limit is 0 since  $p_{\gamma_t}(0) = 0$ )

- and 2.:

$$p_{\gamma_t}(x) = \int_0^x p_{\gamma_t}(t) dt = 0$$
for  $x = 0$  note:  $p_{\gamma_t}(x) \xleftarrow{Laplace}{transform} \frac{1}{s} M_{\gamma_t}(s)$ 

### General combining scheme

$$y = \left(\sum_{n=1}^{N_R} h_n w_n\right) x + \sum_{n=1}^{N_R} w_n n_n$$
$$\gamma_t = \frac{\epsilon_s \left|\sum_{n=1}^{N_R} h_n w_n\right|^2}{\sigma_n^2 \sum_{N=1}^{N_R} |w_n|^2}$$

where  $w_n$  depends on the particular combining scheme.

# 1.2 MRC (Maximum Ratio Combining)

- what weight  $w_n$  maximize  $\gamma_t$ ?
  - Cauchy-Schwarz inequality

$$\left| \sum_{n=1}^{N_R} h_n w_n \right|^2 \le \sum_{n=1}^{N_R} |h_n|^2 \cdot \sum_{n=1}^{N_R} |w_n|^2$$

where equality holds if and only if  $w_n = c \cdot h_n^*$  for some non-zero constant c.

- for  $w_n = h_n^*$ , we obtain

$$\gamma_t = \frac{\epsilon_s}{\sigma_n^2} \cdot \frac{\left(\sum_{n=1}^{N_R} |h_n|^2\right)^2}{\sum_{n=1}^{N_R} |h_n|^2} = \frac{\epsilon_s}{\sigma_n^2} \sum_{n=1}^{N_R} |h_n|^2$$

- $-w_n = h_n^* \forall n$  are the MRC combining weights.
- For performance analysis we assume independent identically distributed (IID) Rayleigh fading

$$\rightarrow \mathcal{E}\{|h_n|^2\} = 1; \quad \bar{\gamma} = \frac{\epsilon_s}{\sigma_n^2}; \quad \gamma_n = \frac{\epsilon_s}{\sigma_n^2}|h_n|^2$$
$$p_\gamma(x) = \frac{1}{\bar{\gamma}}e^{-\frac{x}{\bar{\gamma}}}; \quad x \ge 0$$
$$M_\gamma(s) = \frac{1}{1 + s\bar{\gamma}}$$

• Error rate

$$\gamma_t = \sum_{n=1}^{N_R} \gamma_n$$

 $\rightarrow$  sum of IID random variables (r.v.s.)

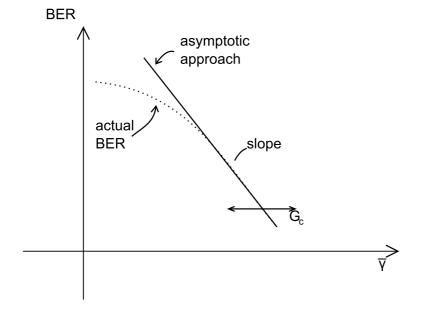
$$M_{\gamma_t}(s) = \left(M_{\gamma}(s)\right)^{N_R} = \frac{1}{(1+s\bar{\gamma})^{N_R}} = \frac{1}{\bar{\gamma}^{N_R}} \cdot \frac{1}{(s+\frac{1}{\bar{\gamma}})^{N_R}}$$

inverse Laplace-transform (from tables)

$$p_{\gamma_t}(x) = \frac{1}{\bar{\gamma}^{N_R}} \cdot \frac{x^{N_R - 1}}{(N_R - 1)!} e^{-\frac{x}{\bar{\gamma}}}; \quad x \ge 0$$

• Direct approach

$$p_e = \int_0^\infty a \cdot Q(\sqrt{ax}) p_{\gamma_t}(x) \ dx = a \left(\frac{1-\mu}{2}\right)^{N_R} \cdot \sum_{n=0}^{N_R-1} \binom{N_R-1+n}{n} \left(\frac{1+\mu}{2}\right)^n$$
 where  $\mu = \sqrt{\frac{b\bar{\gamma}}{2+b\bar{\gamma}}}$ 



• MGF approach

$$p_e = \frac{a}{\pi} \int_0^{\frac{\pi}{2}} M_{\gamma_t} \left( \frac{b}{2 \sin^2 \theta} \right) d\theta$$

$$= \frac{a}{\pi} \int_0^{\frac{\pi}{2}} \frac{1}{\bar{\gamma}^{N_R} \left( \frac{b}{\sin^2 \theta} + \frac{1}{\bar{\gamma}} \right)^{N_R}} d\theta \quad \text{(numerisch berechnen!)}$$

• high SNR:  $\bar{\gamma} \to \infty \Longleftrightarrow \frac{1}{\bar{\gamma}} \to 0$ 

$$\begin{split} p_e &= \frac{a}{\pi} \cdot \frac{1}{\bar{\gamma}^{N_R}} \cdot \left(\frac{2}{b}\right)^{N_R} \int_0^{\frac{\pi}{2}} \sin^{2N_R} \theta \ d\theta \\ \text{(from MGF approach: } \int_0^{\frac{\pi}{2}} \sin^{2N_R} \theta \ d\theta &= \frac{\pi}{2^{N_R+1}} \cdot \binom{2N_R}{N_R} \\ &= \frac{a}{2^{N_R+1} \cdot b^{N_R}} \left(2N_R - N_R\right) \frac{1}{\bar{\gamma}^{N_R}} \quad \text{as } \bar{\gamma} \to \infty \\ &= \frac{1}{G_c \bar{\gamma}} \end{split}$$

where: Diversity gain:  $G_d = N_R$ 

Combining/Coding gain: 
$$G_c = 2b \left(\frac{a}{2} \binom{2N_R}{N_R}\right)^{-\frac{1}{N_R}}$$

- MRC exploits the maximal possible diversity
- Diversity gain is not affected by correlation as the branches are not fully correlated
- Diversity gain depends on fading distribution

### Outage probability

$$P_{out} = \int_0^{\gamma_T} p_{\gamma_t}(x) \ dx = \frac{1}{\bar{\gamma}^{N_R}} \int_0^{\gamma_T} \frac{x^{N_R - 1}}{(N_R - 1)!} e^{-\frac{x}{\bar{\gamma}}} \ dx$$
$$= 1 - e^{-\frac{\gamma_T}{\bar{\gamma}}} \cdot \sum_{n=1}^{N_R} \frac{\left(\frac{\gamma_T}{\bar{\gamma}}\right)^n}{(n-1)!}$$

• Approximation (Taylor series):  $\bar{\gamma} \to \infty$ :  $-e^{-\frac{x}{\bar{\gamma}}} = 1 - \frac{x}{\bar{\gamma}} + O(\frac{1}{\bar{\gamma}})$  where a function f(x) is O(x) if  $\lim_{x \to \infty} \frac{f(x)}{x} = 0$ .

$$\Rightarrow P_{out} = \frac{1}{\gamma^{N_R}} \int_{0}^{\gamma_T} \frac{x^{N_R - 1}}{(N_R - 1)!} \left( 1 - \frac{x}{\bar{\gamma}} + O\left(\frac{1}{\bar{\gamma}}\right) \right)$$

• Diversity and coding gain can also be defined for  $P_{out}$ 

# 1.3 EGC (Equal Gain Combining)

### Combining Weights

- For MRC, both, the amplitudes and phases of the channel gains  $h_n = |h_n|e^{j\varphi_n}$  have to be known (or estimated in practice)
- In EGC it is assumed that only the phases are known and weights  $w_n = e^{-j\varphi_n}$  are used.

$$\Rightarrow \gamma_t = \frac{\mathcal{E}_s}{\sigma_n^2} \frac{\left| \sum_{n=1}^{N_R} |h_n| e^{j\varphi_n} e^{-j\varphi_n} \right|^2}{\sum_{n=1}^{N_R} |e^{-j\varphi_n}|^2} = \frac{\mathcal{E}_s}{\sigma_n^2} \frac{1}{N_R} \left( \sum_{n=1}^{N_R} |h_n| \right)^2$$
$$= \frac{1}{N_R} \left( \sum_{n=1}^{N_R} \sqrt{\gamma_n} \right)^2; \text{ with } \gamma_n = \frac{\mathcal{E}_s}{\sigma_n^2} |h_n|^2$$

### Performance Analysis

- IID case
  - $\Rightarrow \sqrt{\gamma_n}$  is Rayleigh distributed
  - $\Rightarrow$  Exact analysis is much more difficult than for MRC  $\Rightarrow$  see book by Simon & Alouini p.341
- Approximate result

$$P_{e} = \frac{a}{2} \left[ 1 - \sqrt{\frac{2b\bar{\gamma}}{5 + 2b\bar{\gamma}}} \sum_{n=0}^{N_{R}-1} \frac{\binom{2n}{n}}{4^{n} (1 + \frac{2}{5}b\bar{\gamma})^{n}} \right]$$

- high SNR
  - $\Rightarrow$  use high SNR analysis of Wang & Giannakis, 2003
  - $\Rightarrow$  at high SNR, only pdf of  $\gamma_n$  around 0 is relevant for performance

$$\Rightarrow \begin{array}{l} \text{Rayleigh} \\ p_{\gamma}(x) \\ = \\ \frac{1}{\bar{\gamma}} e^{-\frac{x}{\bar{\gamma}}} \end{array} \overset{\text{Taylor Serie}}{=} \frac{1}{\bar{\gamma}} + O\left(\frac{1}{\bar{\gamma}}\right) \text{ as } x \to 0$$

• need pdf  $\gamma_t$ : ( $\gamma_n$  bekannt,  $\rightarrow$  ges.: Wurzel, etc.) (cumulative distribution function of  $\sqrt{\gamma}$  ( $\stackrel{\text{i.i.d}}{=}$   $\sqrt{\gamma_n}$ ) (cdf))

$$\begin{split} &P_{\sqrt{\gamma}}(x) = \Pr \big\{ \sqrt{\gamma} \leq x \big\} = \Pr \big\{ \gamma \leq x^2 \big\} = P_{\gamma}(x^2) = \text{cdf of } \gamma \\ &\to p_{\sqrt{\gamma}}(x) = \frac{d}{dx} P_{\sqrt{\gamma}}(x) = 2x \cdot p_{\gamma}(x^2) = \frac{2x}{\bar{\gamma}} + O \Big( \frac{1}{\bar{\gamma}} \Big) \end{split}$$

• Laplace Transformation to MGF

• inverse Laplace Transform

$$\begin{split} p_{\sqrt{\gamma_{t}}}(x) &= \mathcal{L}^{-1}\Big\{M_{\sqrt{\gamma_{t}}}(s)\Big\} = \left(\frac{2N_{R}}{\bar{\gamma}}\right)^{N_{R}} \cdot \frac{x^{2N_{R}-1}}{(2N_{R}-1)!} + O\left(\frac{1}{\bar{\gamma}^{N_{R}}}\right) \\ P_{\gamma_{t}}(x) &= \Pr\{\gamma_{t} \leq x\} = \Pr\{\sqrt{\gamma_{t}} \leq \sqrt{x}\} = P_{\sqrt{\gamma_{t}}}(\sqrt{x}) \to \operatorname{cdf of } \sqrt{\gamma_{t}} \\ p_{\gamma_{t}}(x) &= \frac{d}{dx}P_{\gamma_{t}}(x) = \frac{1}{2\sqrt{x}} \cdot p_{\gamma_{t}}(\sqrt{x}) = \frac{1}{2}\left(\frac{2N_{R}}{\bar{\gamma}}\right)^{N_{R}} \cdot \frac{x^{N_{R}-1}}{(2N_{R}-1)!} + O\left(\bar{\gamma}^{-N_{R}}\right) \\ \to M_{\gamma_{t}}(s) &= \mathcal{L}\{p_{\gamma_{t}}(x)\} = \frac{1}{2}\left(\frac{2N_{R}}{\bar{\gamma}}\right)^{N_{R}} \cdot \frac{(N_{R}-1)!}{(2N_{R}-1)!} + O\left(\bar{\gamma}^{-N_{R}}\right) \end{split}$$

• Error Probability:

$$P_{e} = \frac{a}{\pi} \int_{0}^{\frac{\pi}{2}} M_{\gamma_{t}} \left(\frac{b}{2 \sin^{2}(\theta)}\right) d\theta$$

$$= \frac{a}{\pi} \frac{1}{2} \left(\frac{2N_{R}}{\bar{\gamma}}\right)^{N_{R}} \frac{(N_{R} - 1)!}{(2N_{R} - 1)!} \frac{2^{N_{R}}}{b^{N_{R}}} \int_{\frac{\pi}{2^{2N_{R} + 1}}}^{\frac{\pi}{2}} \sin^{2N_{R}}(\theta) d\theta + O\left(\frac{1}{\bar{\gamma}^{N_{R}}}\right)$$

$$= \frac{aN_{R}^{N_{R}}}{2b^{N_{R}}N_{R}!} \frac{1}{\bar{\gamma}^{N_{R}}} + O\left(\frac{1}{\bar{\gamma}^{N_{R}}}\right) \stackrel{!}{=} \left(\frac{1}{G_{c}}\right)^{G_{d}}$$

$$\implies \text{Diversity gain: } G_{d} = N_{R}$$

$$\implies \text{Combining gain: } G_{c} = \frac{b}{N_{R}} \left(\frac{2N_{R}!}{a}\right)^{\frac{1}{N_{R}}}$$

vergleiche auch Blatt mit Kurven  $\overline{\underline{III}}$  und  $\overline{\underline{IV}}$ 

A similar asymptotic analysis can be conducted for the outage probability.

# 1.4 SC (Selection Combining)

#### Combining weights

- only the strongest branch is chosen
- strongest branch:  $\hat{n} = \underset{n}{\operatorname{argmax}} \gamma_n \longrightarrow \gamma_t = \gamma_{\hat{n}}$
- $\bullet\,$  only on RF receiver chain required  $\to\,$  saves hardware complexity

#### Performance analysis

• cdf of:  $\gamma_t$ 

$$P_{\gamma_t}(x) = \Pr\{\gamma_{\hat{n}} \le x\} = \Pr\{\gamma_1 \le x \cap \gamma_2 \le x \cap \dots \gamma_{N_R} \le x\}$$

$$\stackrel{(IID)}{=} \left(\Pr\{\gamma_n \le x\}\right)^{N_R} = \left(P_{\gamma}(x)\right)^{N_R}$$

• pdf:

$$\begin{split} p_{\gamma_t}(x) &= \frac{d}{dx} P_{\gamma_t}(x) = N_R \big( P_{\gamma}(x) \big)^{N_R - 1} \cdot p_{\gamma}(x) \\ \text{where:} \qquad p_{\gamma_t}(x) &= \frac{1}{\bar{\gamma}} e^{-\frac{x}{\bar{\gamma}}}; \quad x \geq 0 \\ P_{\gamma}(x) &= \int_0^x p_{\gamma}(x) \; dx = 1 - e^{-\frac{x}{\bar{\gamma}}}; \quad x \geq 0 \\ &\to p_{\gamma_t}(x) = \frac{N_R}{\bar{\gamma}} \big( 1 - e^{-\frac{x}{\bar{\gamma}}} \big)^{N_R - 1} e^{-\frac{x}{\bar{\gamma}}}; \quad x \geq 0 \end{split}$$

### Error probability

- $\bullet$  direct approach  $\rightarrow$  closed-form solution possible
- MGF approach
  - Binomial expansion

$$p_{\gamma_t}(x) = \frac{N_R}{\bar{\gamma}} e^{-\frac{x}{\bar{\gamma}}} \sum_{n=0}^{N_R - 1} \binom{N_R - 1}{n} 1^{N_R - 1 - n} \left( -e^{-\frac{x}{\bar{\gamma}}} \right)^n$$
$$= \frac{N_R}{\bar{\gamma}} \sum_{n=0}^{N_R - 1} \binom{N_R - 1}{n} \cdot (-1)^n e^{-\frac{x(n+1)}{\bar{\gamma}}}; \quad x \ge 0$$

- MGF

$$M_{\gamma_t}(s) = \frac{N_R}{\bar{\gamma}} \sum_{n=0}^{N_R - 1} {N_R - 1 \choose n} (-1)^n \frac{1}{s + \frac{n+1}{\bar{\gamma}}}$$

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$$P_e = \frac{a}{\pi} \int_0^{\frac{\pi}{2}} M_{\gamma_t} \left( \frac{b}{2 \sin^2 \theta} \right) d\theta = \frac{a N_R}{\pi \bar{\gamma}} \sum_{n=0}^{N_R - 1} \binom{N_R - 1}{n} (-1)^n \int_0^{\frac{\pi}{2}} \frac{d\theta}{\frac{b}{2 \sin^2 \theta} + \frac{n+1}{\bar{\gamma}}}$$

 $\rightarrow$  can be evaluated numerically

– high SNR approach  $\Rightarrow \bar{\gamma} \to \infty$ 

$$p_{\gamma_t} = \frac{N_R}{\bar{\gamma}} \left[ 1 - \exp\left(-\frac{x}{\bar{\gamma}}\right) \right]^{N_R - 1} \exp\left(-\frac{x}{\bar{\gamma}}\right)$$

$$\stackrel{\bar{\gamma} \to \infty}{=} \frac{N_R}{\bar{\gamma}} \left[ 1 - \left(1 - \frac{x}{\bar{\gamma}} + O\left(\bar{\gamma}^{-1}\right)\right) \right]^{N_R - 1} \left(1 - \frac{x}{\bar{\gamma}} + O\left(\bar{\gamma}^{-1}\right)\right) i$$

$$= \frac{N_R}{\bar{\gamma}^{N_R}} x^{N_R - 1} + o\left(\bar{\gamma}^{-N_R}\right)$$

- MGF:

$$\begin{split} M_{\gamma_t}(s) &= \frac{N_R}{\bar{\gamma}^{N_R}} \frac{(N_R - 1)!}{s^{N_R}} + O\left(\bar{\gamma}^{-N_R}\right) \\ \left[ \to P_e &= \frac{a}{\pi} \int\limits_0^{\frac{\pi}{2}} M_{\gamma_t} \left(\frac{b}{2\sin^2(\theta)}\right) \mathrm{d}\theta \right] \\ &= \frac{a(2N_R)!}{b^{N_R} 2^{N_R + 1} N_R!} \frac{1}{\bar{\gamma}^{N_R}} + O(\bar{\gamma}^{-N_R}) \end{split}$$

 $\Longrightarrow$  Diversity gain:  $G_d = N_R$ 

$$\implies$$
 Combining gain:  $G_c = 2b \left( \frac{2N_R!}{a(2N_R)!} \right)^{\frac{1}{N_R}}$ 

### - Outage Probability

$$\begin{split} P_{out} &= \Pr\{\gamma_{\hat{n}} \leq \gamma_T\} = P_{\gamma_{\hat{n}}}(\gamma_T) = \left[1 - \exp\left(-\frac{\gamma_T}{\bar{\gamma}}\right)\right]^{N_R} \\ \text{high SNR: } P_{out} &= \left(\frac{\gamma_T}{\bar{\gamma}}\right)^{N_R} + O\left(\bar{\gamma}^{-N_R}\right) \end{split}$$