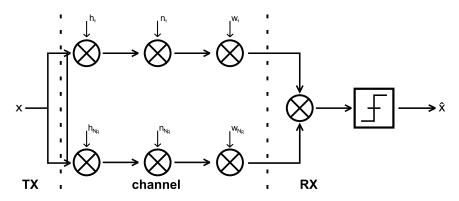
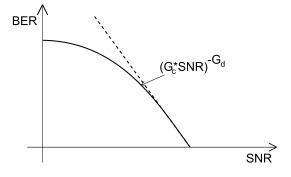
1 SIMO Systems

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 γ_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:

$$Q(x) = \frac{1}{\pi} \int_{0}^{\frac{\pi}{2}} e^{-\frac{x^{2}}{2sin^{2}\theta}} d\theta$$

- → 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_{γ_t}
- $-M_{\gamma_t}(-s)$ is the so called Moment Generation Function (MGF) of p_{γ_t}
- Here, we will also refer to $M_{\gamma_t}(s)$ as MGF
- $M_{\gamma_t}(s)$ is sometimes easier to obtain than p_{γ_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 γ_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



- 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 γ_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.
- \bullet For performance analysis we assume independent identically distributed (IID) Rayleigh fading

• Error rate

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

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

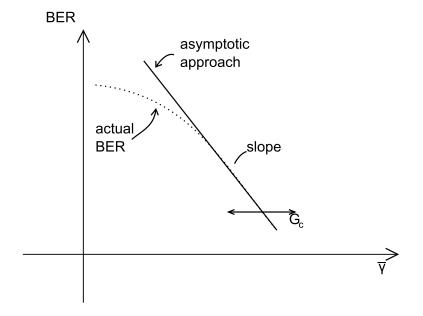
$$M_{\gamma_t}(s) = \left(M_{\gamma}(s)\right)^{N_R} = \frac{1}{(1+s\bar{\gamma})^{N_R}} = \frac{1}{\bar{\gamma}^{N_R}} \cdot \frac{1}{(s+\frac{1}{\bar{z}})^{N_R}}$$

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

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



• MGF approach

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

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

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

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
 - \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 γ_n around 0 is relevant for performance

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

• need pdf γ_t : (γ_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

$$\begin{split} & \to M_{\sqrt{\gamma}}(s) = \mathcal{L}\big\{p_{\sqrt{\gamma}}(x)\big\} = \frac{2}{\bar{\gamma}} \cdot \frac{1}{s^2} + O\big(\frac{1}{\bar{\gamma}}\big) \\ & \sqrt{\gamma_t} = \sum_{n=1}^{N_R} \frac{\sqrt{\gamma_n}}{N_R} \\ & M_{\sqrt{\gamma_t}}(s) = \mathcal{E}\Big\{\exp(-s\sqrt{\gamma_t})\Big\} = \mathcal{E}\Big\{\exp(-\frac{s}{\sqrt{N_R}} \cdot \sum_{n=1}^{N_R} \sqrt{\gamma_n})\Big\} = \Big(\mathcal{E}\Big\{\exp(-\frac{s}{\sqrt{N_R}} \cdot \sqrt{\gamma_n}\Big\}\Big)^{N_R} \\ & = \Big(M_{\sqrt{\gamma}}\big(\frac{s}{\sqrt{N_R}}\big)\Big)^{N_R} = \Big(\frac{2}{\bar{\gamma}} \cdot \frac{N_R}{s^2}\Big)^{N_R} + O\Big(\frac{1}{\bar{\gamma}^{N_R}}\Big) \end{split}$$

• inverse Laplace Transform

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

• Error Probability:

$$\begin{split} P_e &= \frac{a}{\pi} \int\limits_0^{\frac{\pi}{2}} M_{\gamma_t} \left(\frac{b}{2 \sin^2(\theta)} \right) \mathrm{d}\theta \\ &= \frac{a}{\pi} \frac{1}{2} \left(\frac{2N_R}{\bar{\gamma}} \right)^{N_R} \frac{(N_R - 1)!}{(2N_R - 1)!} \frac{2^{N_R}}{b^{N_R}} \int\limits_{\frac{\pi}{2^{2N_R + 1}}}^{\frac{\pi}{2}} \sin^{2N_R}(\theta) \mathrm{d}\theta + O\left(\frac{1}{\bar{\gamma}^{N_R}} \right) \\ &= \frac{aN_R^{N_R}}{2b^{N_R}N_R!} \frac{1}{\bar{\gamma}^{N_R}} + O\left(\frac{1}{\bar{\gamma}^{N_R}} \right) \stackrel{!}{=} \left(\frac{1}{G_c} \right)^{G_d} \\ \Longrightarrow \text{Diversity gain: } G_d = N_R \\ \Longrightarrow \text{Combining gain: } G_c = \frac{b}{N_R} \left(\frac{2N_R!}{a} \right)^{\frac{1}{N_R}} \end{split}$$

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

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: $\hat{n} = \underset{n}{\operatorname{argmax}} \gamma_n \longrightarrow \gamma_t = \gamma_{\hat{n}}$
- ullet only on RF receiver chain required o saves hardware complexity

Performance analysis

• cdf of: γ_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

$$\begin{split} p_{\gamma_t}(x) &= \frac{N_R}{\bar{\gamma}} e^{-\frac{x}{\bar{\gamma}}} \sum_{n=0}^{N_R-1} \binom{N_R-1}{n} 1^{N_R-1-n} \Big(-e^{-\frac{x}{\bar{\gamma}}} \Big)^n \\ &= \frac{N_R}{\bar{\gamma}} \sum_{n=0}^{N_R-1} \binom{N_R-1}{n} \cdot (-1)^n e^{-\frac{x(n+1)}{\bar{\gamma}}}; \quad x \geq 0 \end{split}$$

- MGF

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

_

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

 \rightarrow can be evaluated numerically

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

$$\begin{split} p_{\gamma_t} &= \frac{N_R}{\bar{\gamma}} \left[1 - \exp\left(-\frac{x}{\bar{\gamma}}\right) \right]^{N_R - 1} \exp\left(-\frac{x}{\bar{\gamma}}\right) \\ &\stackrel{\bar{\gamma} \to \infty}{=} \frac{N_R}{\bar{\gamma}} \left[1 - \left(1 - \frac{x}{\bar{\gamma}} + O\left(\bar{\gamma}^{-1}\right)\right) \right]^{N_R - 1} \left(1 - \frac{x}{\bar{\gamma}} + O\left(\bar{\gamma}^{-1}\right)\right) i \\ &= \frac{N_R}{\bar{\gamma}^{N_R}} x^{N_R - 1} + o\left(\bar{\gamma}^{-N_R}\right) \end{split}$$

- MGF:

$$\begin{split} M_{\gamma_t}(s) &= \frac{N_R}{\bar{\gamma}^{N_R}} \frac{(N_R - 1)!}{s^{N_R}} + O\left(\bar{\gamma}^{-N_R}\right) \\ \left[\to P_e &= \frac{a}{\pi} \int_0^{\frac{\pi}{2}} M_{\gamma_t} \left(\frac{b}{2\sin^2(\theta)}\right) \mathrm{d}\theta \right] \\ &= \frac{a(2N_R)!}{b^{N_R} 2^{N_R + 1} N_R!} \frac{1}{\bar{\gamma}^{N_R}} + O(\bar{\gamma}^{-N_R}) \end{split}$$

 \implies 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

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

1.5 Comparison

- Diversity Gain: MRC, EGC and SC all achieve the maximum possible diversity gain of $G_d=N_R$
- Combining Gain:
 The combining gains of MRC, EGC and SC are different
 - MRC/EGC:

$$\frac{G_C^{EGC}}{G_C^{MRC}} = \frac{\frac{1}{2b} \left(\frac{a}{2} {2N_R \choose N_R}\right)^{\frac{1}{N_R}}}{\frac{N_R}{b} \left(\frac{a}{2} \frac{1}{N_R!}\right)^{\frac{1}{N_R}}} = \frac{\left[(2N_R)!\right]^{\frac{1}{N_R}}}{2N_R(N_R)^{\frac{1}{N_R}}} \le 1$$

(independent of a or b which are modulation parameters, only depends on number of antennas)

$$N_R \gg 1: \qquad N_R! \approx \sqrt{2\pi} e^{-N_R} N_R^{N_R + \frac{1}{2}} \qquad (Stirling)$$

$$\frac{G_C^{EGC}}{G_C^{MRC}} \bigg|_{N_R \gg 1} = \frac{\left(\sqrt{2\pi} e^{-2N_R} (2N_R)^{2N_R + \frac{1}{2}}\right)^{\frac{1}{N_R}}}{2N_R \left(\sqrt{2\pi} e^{-N_R} N_R^{N_R + \frac{1}{2}}\right)^{\frac{1}{N_R}}} = \frac{2 \cdot 2^{\frac{1}{2N_R}}}{2} \xrightarrow{N_R \to \infty} \frac{2}{e} \equiv -1.3 \text{dB}$$

- MRC/SC:

$$\begin{split} \frac{G_C^{SC}}{G_C^{MRC}} &= \frac{2b \left(\frac{a}{2} \binom{2N_R}{N_R}\right)^{\frac{1}{N_R}}}{2b \left(\frac{a}{2} \frac{(2N_R)!}{N_R!}\right)^{\frac{1}{N_R}}} = \frac{1}{(N_R!)^{\frac{1}{N_R}}} \leq 1 \\ \frac{G_C^{SC}}{G_C^{MRC}} \bigg|_{N_R \gg 1} &= \frac{1}{\sqrt{2\pi^{\frac{1}{N_R}}}e^{-1}N_R^{\frac{1+\frac{1}{2N_R}}{2N_R}}} N_R \xrightarrow{\rightarrow} \infty \frac{e}{N_R} \end{split}$$

 \rightarrow loss increases with N_R

2 MISO Systems

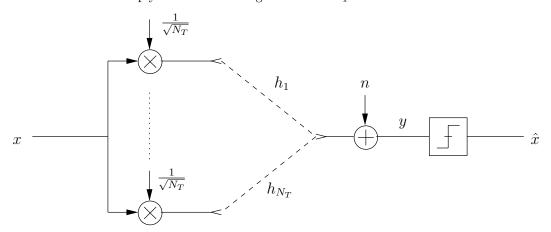
Remarks

• Similar to SIMO systems, in MISO systems only coding and diversity gains can be obtained.

- To realize these gains, a careful transmitter design is necessary
- ullet System design depends on whether or not channel state information (CSI) is available at transmitter

2.1 Naive Approach

• Assume we simply send the same signal over all N_T transmit antennas



- Transmit power: $\mathcal{E}\left\{\left|\frac{1}{\sqrt{N_T}}x\right|^2+,\ldots,\left|\frac{1}{\sqrt{N_T}}x\right|^2\right\}=\mathcal{E}\left\{N_T\frac{1}{N_T}|x|^2\right\}=\mathcal{E}_s$
- Received signal: $y = \frac{1}{\sqrt{N_T}} \sum_{n=1}^{N_T} h_n \cdot x + n$
- Rayleigh fading: h_n are zero mean complex gaussian random variables $\to h$ is also zero mean complex gaussian
- i.i.d.:

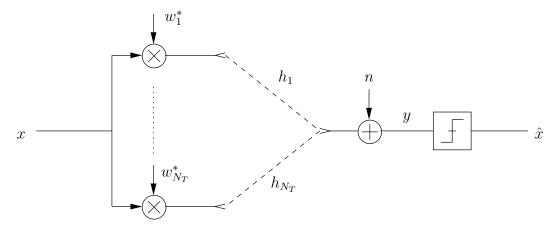
$$- \mathcal{E}\{|h_n|^2\} = 1 \ \forall n$$

$$- \mathcal{E}\{|h|^2\} = \frac{1}{N_T} \mathcal{E}\left\{ \left| \sum_{n=1}^{N_T} h_n \right|^2 \right\} = \frac{1}{N_T} \mathcal{E}\left\{ \sum_{n=1}^{N_T} |h_n|^2 \right\} = 1$$

- statistical properties of h are independent of N_T
- the multiple transmit antennas have no benefit at all
- more sophisticated transmitter designs necessary

2.2 Full CSI Available at the Transmitter

- $h_n, n \in \{1, \ldots, N_T\}$ is known at the transmitter
- Perform "precoding" (beamforming) with coefficients w_n



- Transmit Power: Two constraints maybe considered
 - Average transmit power constraint

$$P_{av} = \mathcal{E}\left\{\sum_{n=1}^{N_T} |w_n^* x|^2\right\} = \sum_{n=1}^{N_T} |w_n|^2 \underbrace{\mathcal{E}\{|x|^2\}}_{\mathcal{E}_s} = \mathcal{E}_s \Rightarrow \sum_{n=1}^{N_T} |w_n|^2 = 1$$

- Power constraint for each transmit antenna

$$\rightarrow |w_n| = \frac{1}{\sqrt{N_T}} \longrightarrow P_{av} = \mathcal{E}_s$$

• Received signal: $y = \underbrace{\sum_{n=1}^{N_T} w_n^* h_n}_{h} x + n$ (equivalent SISO channel)

2.2.1 Maximum Ratio Transmission (MRT)

ullet we have only the average power constraint: $\sum\limits_{n=1}^{N_T} |w_n| = 1$

• SNR:
$$\gamma_t = \frac{\epsilon_s |h|^2}{\sigma_n^2} = \frac{\epsilon_s |\sum_{n=1}^{N_T} w_n^* \cdot h_n|^2}{\sigma_n^2}$$

- Maximize SNR under constraint $\sum_{n=1}^{N_T} |w_n|^2 = 1$
- \bullet constraint optimization problem \to Lagrange method

$$L = \frac{\epsilon_s}{\sigma_n^2} \left| \sum_{n=1}^{N_T} w_n^* \cdot h_n \right|^2 + \lambda \left(\sum_{n=1}^{N_T} |w_n|^2 - 1 \right); \quad \text{where: } \lambda = \text{Lagrange Multiplier}$$

 \Rightarrow Wirtinger Kalkül: treat z and z^* as independent variables for differentiation:

$$\frac{\partial z^*}{\partial z} = 0; \quad \frac{\partial |z|^2}{\partial z} = \frac{\partial z \cdot z^*}{\partial z} = z$$

$$\frac{\partial x^2}{\partial x} = 2x; \quad \frac{\partial (z^*)^2}{\partial z^*} = 2 \cdot z^*; \frac{\partial |z|^2}{\partial z} = z$$

$$\frac{\partial L}{\partial w_m^*} = \frac{\epsilon_s}{\sigma_n^2} \Big(\sum_{n=1}^{N_T} w_n^* \cdot h_n\Big)^* h_m + \lambda w_m$$

$$\to w_m = \frac{\epsilon_s}{\sigma_n^2 \cdot \lambda} \Big(\sum_{n=1}^{N_T} w_n^* h_n\Big)^* h_m$$

$$\text{const., independent of } m := c$$

$$\to w_m = c \cdot h_m$$

$$\to \sum_{n=1}^{N_T} |w_n|^2 = 1 \to c^2 = \frac{1}{\sum_{n=1}^{N_T} |h_n|^2}$$

$$\to w_n = \frac{h_n}{\sqrt{\sum_{n=1}^{N_T} |h_n|^2}} \equiv \text{MRT gains}$$

$$\text{SNR} = \frac{\epsilon_s}{|w_n|^2} \Big|\sum_{n=1}^{N_T} \frac{|h_n|^2}{|h_n|^2} \Big|^2 = \frac{\epsilon_s}{|w_n|^2} \sum_{n=1}^{N_T} |h_n|^2$$

$$\to \text{SNR} = \frac{\epsilon_s}{\sigma_n^2} \left| \sum_{n=1}^{N_T} \frac{|h_n|^2}{\sqrt{\sum_{n=1}^{N_T} |h_m|^2}} \right|^2 = \frac{\epsilon_s}{\sigma_n^2} \sum_{n=1}^{N_T} |h_n|^2$$

- \Rightarrow same SNR as for maximum ration combining (MRC)
- \Rightarrow MRT with $N_R t_x$ antennas achieves the same performance as MRC with $N_T r_x$ antennas
- \Rightarrow MRT/MIMO can be extended to $N_T \times N_R$ MIMO systems
 - $\rightarrow\,$ has the same performance as MRC with $N_T N_R \; r_x$ antennas and one t_x antenna

Equal Gain Transmission (EGT)

- we employ gains: $w_n = \frac{1}{\sqrt{N_T}} \cdot \frac{h_n}{|h_m|} \to |w_n| = \frac{1}{\sqrt{N_T}}$
- SNR:

$$\begin{split} \gamma_t &= \frac{\epsilon_s}{\sigma_n^2} \Big| \sum_{n=1}^{N_T} w_n^* h_n \Big|^2 \\ &= \frac{\epsilon_s}{\sigma_n^2} \Big| \sum_{n=1}^{N_T} \frac{1}{\sqrt{N_T} \cdot \frac{|h_n|^2}{|h_n|}} \Big|^2 = \frac{1}{N_T} \cdot \frac{\epsilon_s}{\sigma_n^2} \Big| \sum_{n=1}^{N_T} |h_n| \Big|^2 \\ \gamma_n &= \frac{\epsilon_s}{\sigma_n^2} |h_n|^2 \\ \text{same SNR as for EGC} &\to \gamma_t = \frac{1}{N_T} \Big| \sum_{n=1}^{N_T} N_T \sqrt{\gamma_n} \Big|^2 \end{split}$$

 \rightarrow EGC with N_T t_x antennas achieves the same performance as EGC with N_T r_x antennas

Transmit Antennas Selection

• select antenna with maximum channel gain for transmission:

$$w_n = \begin{cases} \frac{h_n}{|h_n|}, & \text{if } n = \hat{n} \\ 0, & \text{otherwise} \end{cases} \text{ where } \hat{n} = \underset{n}{\operatorname{argmax}} |h_n|$$

 \bullet antenna selection with N_T t_x antennas achieves the same performance as Selection Combining with N_T r_x antennas