

2G (~ 1990's)		
Gen.	Technology	≈ Data Rate
2G	GSM	10 Kbps / user
2G	CDMA	10 Kbps
2.5G	GPRS	≈ 50 Kbps
2.5G	EDGE	≈ 200 Kbps
GSM : Global System for Mobile CDMA : Code Division for Multiple Access GPRS : General Packet Radio Service EDGE : Enhanced Data for GSM Evolution		

3G (~ 2000's)		
Gen.	Technology	≈ Data Rate
3G	WCDMA/UMTS	384 Kbps
3G	CDMA 2000	384 Kbps
3.5G	HSDPA / HSUPA	5 – 30 Mbps
3.5G	1 EVDO (Rev A,B,C)	5 – 30 Mbps
WCMD : Wideband CDMA UMTS : Universal Mobile Telecomm. Standard HSDPA : High Speed Downlink Packet Access EVDO : Evolution Data Optimized		

4G (~ 2010's)		
Gen.	Technology	≈ Data Rate
4G	LTE	100 – 200 Mbps
4G	WiMax 2000	100 Kbps
4G	LTE-Adv	0.5 – 1 Gbps

LTE : Long Term Evolution
WiMax : Worldwide Interoperability for Microwave Access

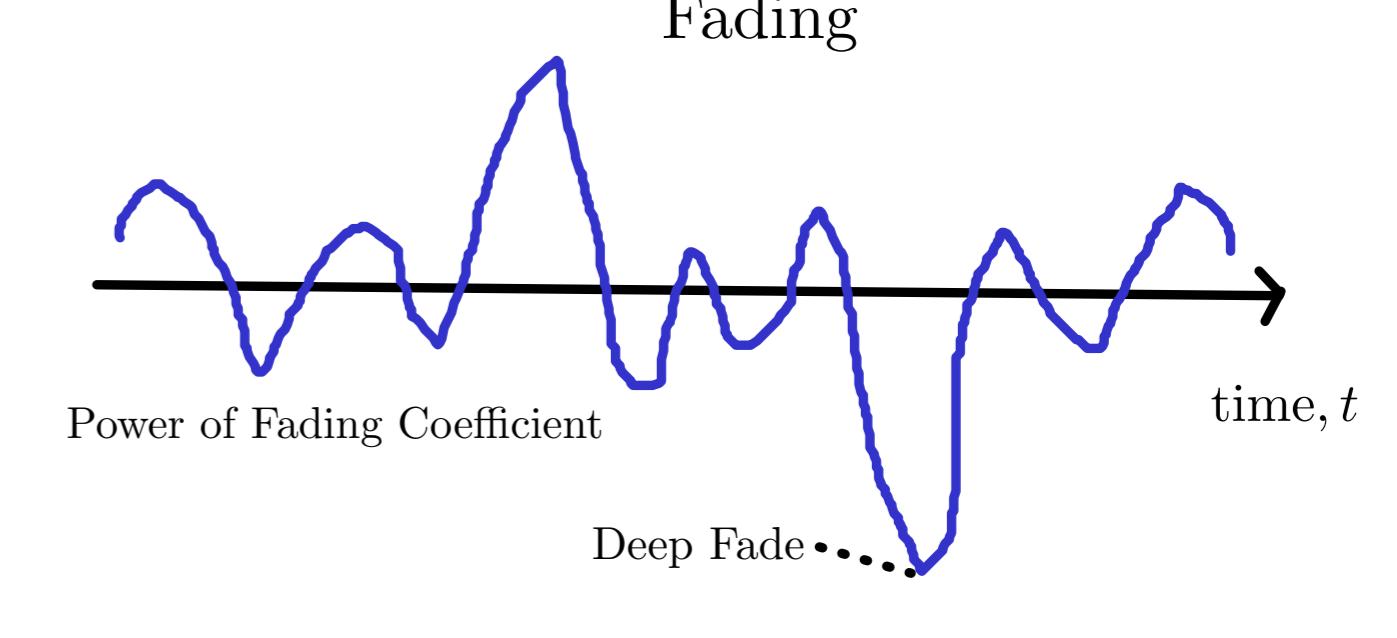
Evolution of Applications in 3G/4G			
Gen.	≈ Data Rate	Enablers	Applications
2G / 2.5G	10 – 100 Kbps		Voice + Basic data
3G / 3.5G	300 Kbps – 30 Mbps	CDMA, MIMO	Voice, High speed Data, Video calling
4G	> 100 Mbps	MIMO, OFDM	Online Gaming, HDTV

→ Increase in data rates and reliability made possible by
MIMO: Multiple Input Multiple Output
CDMA: Code Division for Multiple Access
OFDM: Orthogonal Frequency Division Multiplexing

Received Signal			
i) Received signal $y_p(t)$ = sum of the various multipath components is			
$y_p(t) = \sum_{i=0}^{L-1} \Re \left\{ a_i S(t - \tau_i) e^{j2\pi f_c(t - \tau_i)} \right\}$	$= \Re \left\{ \left(\sum_{i=0}^{L-1} a_i S(t - \tau_i) e^{-j2\pi f_c \tau_i} \right) e^{j2\pi f_c t} \right\}$, where		
	Complex Baseband Received Signal	Complex Phase Factor	

- ii) Under **Narrowband** assumption i.e., $f_m \leq f_c$, we have
 $S(t - \tau_i) \approx S(t)$
 $\Rightarrow y(t) = h \times S(t)$, where **channel coefficient** is
 $h \triangleq \left(\sum_{i=0}^{L-1} a_i e^{-j2\pi f_c \tau_i} \right)$
depends on the **attenuation** and **delay** of different multipath components

Example: h with $L = 2$			
i)	Then $h = a_0 e^{-j2\pi f_c \tau_0} + a_1 e^{-j2\pi f_c \tau_1}$		
ii)	Let $a_0 = a_1 = 1$ and $\tau_0 = 0, \tau_1 = \frac{1}{2f_c}$, then $h = 1e^0 + 1e^{-j\pi} = 0$ (Destructive interference) $y(t) = h \times s(t) = 0$		
iii)	Let $a_0 = a_1 = 1$ and $\tau_0 = 0, \tau_1 = \frac{1}{f_c}$, then $h = 1e^0 + 1e^{-j2\pi} = 2$ (Constructive interference) $y(t) = h \times s(t) = 2s(t)$ (Enhanced signal amplitude)		
iv)	The power of channel coefficient h varies with time		



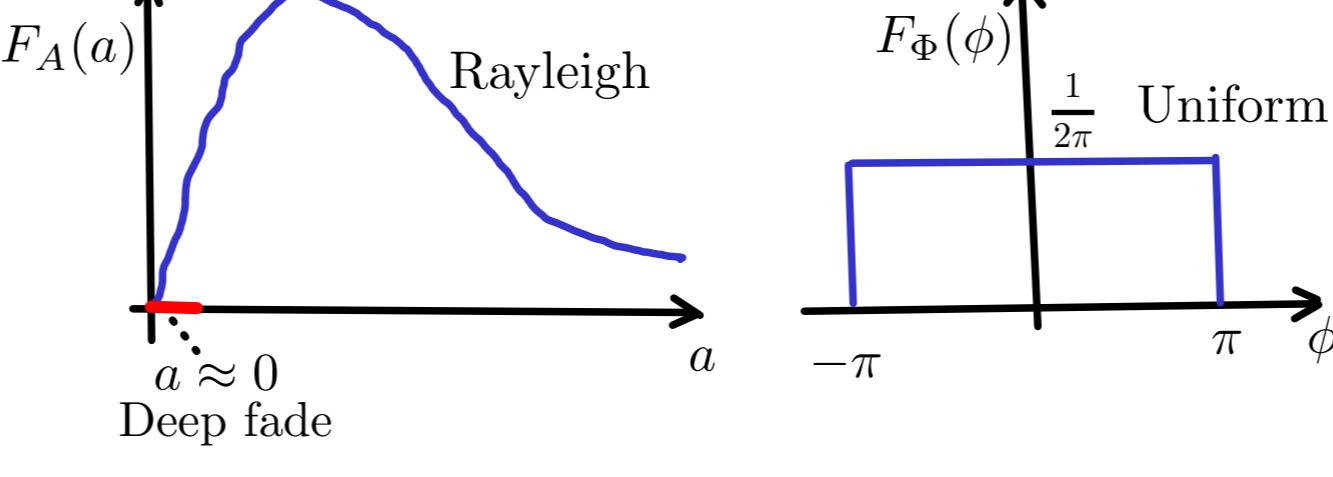
- v) This **Fading** process causes the received power to vary and is a key aspect of wireless channels

Example:			
i)	To compute the probability that the channel attenuation is worse than 20 dB		
ii)	We have channel gain = a^2 $\Rightarrow 10 \log_{10} a^2 < -20 \text{ dB}$, $\Rightarrow a < 0.1$, then $Pr(a < 0.1) = \int_0^{0.1} 2ae^{-a^2} da$ $= -e^{-a^2} \Big _0^{0.1} = 1 - e^{-0.01} \approx 0.01$		
iii)	Using $1 - e^{-x} \approx x$, for small x		

Challenges for Wireless		
i) Presence of scatterers lead to multipath propagation		
→ Leads to superposition of multiple signals		
→ Further leads to constructive or destructive interference		
ii) Constructive interf. amplifies received signal amplitude		
→ Destructive interf. attenuates received signal amplitude		
iii) Develop a model for multipath propagation		

Fading Channel Coefficient (Real and Img. Model)		
i)	We have real and imaginary parts $h = x + jy$, where $x = \sum_{i=0}^{L-1} a_i \cos(2\pi f_c \tau_i)$ and $y = -\sum_{i=0}^{L-1} a_i \sin(2\pi f_c \tau_i)$	
ii)	Note a_i, τ_i are random variables and x, y are sum of large number of random variables \Rightarrow by CLT x, y are assumed to be Gaussian distributed	
iii)	Assuming x, y to be i.i.d. $\sim \mathcal{N}(0, \frac{1}{2})$, we have $F_{X,Y}(x, y) = F_X(x)F_Y(y)$ $= \frac{1}{\sqrt{\pi}} e^{-x^2} \frac{1}{\sqrt{\pi}} e^{-y^2}$ $= \frac{1}{\sqrt{\pi}} e^{-(x^2+y^2)}$	

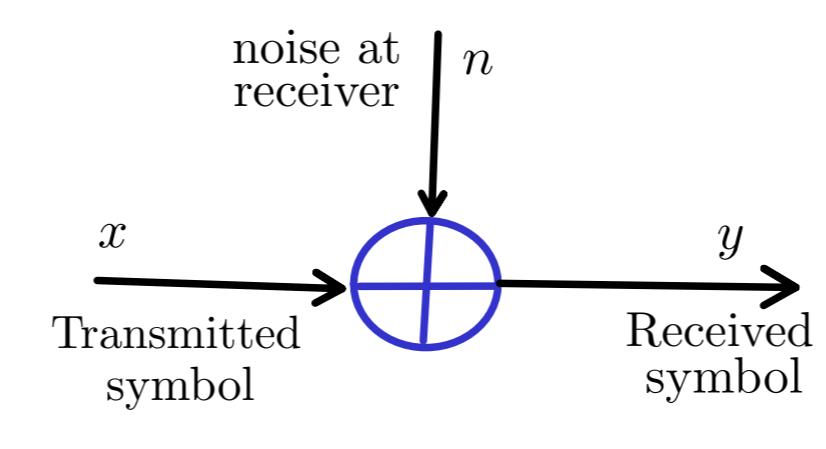
Fading Channel Coefficient (Exponential Model)			
i)	We have $h = x + jy = ae^{j\phi}$, where Amplitude $a = \sqrt{x^2 + y^2}$ and phase $\phi = \tan^{-1} \frac{y}{x}$		
ii)	Then joint distribution $F_{A,\Phi}(a, \phi) = \frac{1}{\pi} e^{-(x^2+y^2)} J_{XY} $ where Det. of Jacobian $ J_{XY} = \begin{vmatrix} \frac{\partial x}{\partial a} & \frac{\partial y}{\partial a} \\ \frac{\partial x}{\partial \phi} & \frac{\partial y}{\partial \phi} \end{vmatrix} = \begin{vmatrix} \cos \phi & \sin \phi \\ -a \sin \phi & a \cos \phi \end{vmatrix} = a$. $\therefore F_{A,\Phi}(a, \phi) = \frac{a}{\pi} e^{-a^2}$		
iii)	Marginal distribution of amplitude A is $F_A(a) = \int_{-\pi}^{\pi} F_{A,\Phi}(a, \phi) d\phi$ $= \frac{a}{\pi} e^{-a^2} \int_{-\pi}^{\pi} d\phi = 2ae^{-a^2} \text{ for } 0 \leq a \leq \infty$ is a Rayleigh distribution , and as result h is also termed as rayleigh fading channel.		
iv)	Marginal distribution of phase Φ is $F_\Phi(\phi) = \int_{-\infty}^{\infty} F_{A,\Phi}(a, \phi) da$ $= \frac{1}{2\pi} \int_0^{\infty} 2ae^{-a^2} da = \frac{1}{2\pi} \text{ for } -\pi < \phi \leq \pi$ is a Uniform distribution .		
v)	Note that $F_{A,\Phi}(a, \phi) = F_A(a)F_\Phi(\phi)$, $\Rightarrow A$ and Φ are independent RVs.		



Performance Analysis (BER)

Wireline Communication system (AWGN)

i) Consider Additive White Gaussian Noise wireline system



we have, $y = x + n$
where n is Gaussian noise with $F_N(n) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{n^2}{2\sigma^2}}$

ii) Consider $x = -\sqrt{P}$, then at receiver $y = x + n = -\sqrt{P} + n$
 \Rightarrow **error occurs** if $y \geq 0$, $\Rightarrow n \geq \sqrt{P}$
 $P_e = Pr(n > \sqrt{P})$
 $= \int_{\sqrt{P}}^{\infty} \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{n^2}{2\sigma^2}} dn = \int_{\sqrt{P}}^{\infty} \frac{1}{\sqrt{\sigma^2/2}} e^{-\frac{t^2}{2}} dt$
 $\Rightarrow P_e = Pr(N(0, 1) \geq \sqrt{\frac{P}{\sigma^2}}) = Q\left(\sqrt{\frac{P}{\sigma^2}}\right) = Q\left(\sqrt{SNR}\right)$ ■

Numerical Example: Exact BER

i) To compute the BER at $SNR = 10dB$
 $\rightarrow 10 \log_{10} SNR = 10dB, \Rightarrow SNR = 10$
 $\rightarrow P_e = Q\left(\sqrt{SNR}\right) = Q\left(\sqrt{10}\right) \approx 7.82 \times 10^{-4}$

Approximate BER

i) Using the **approximation** $Q(x) \leq \frac{1}{2} e^{-\frac{x^2}{2}}$,
 $\rightarrow P_e = Q\left(\sqrt{SNR}\right) \leq \frac{1}{2} e^{-\frac{SNR}{2}}$ ■

Numerical Example: Approx SNR

i) To compute the SNR required for $P_e = 10^{-6}$
 $\rightarrow \frac{1}{2} e^{-\frac{SNR}{2}} = 10^{-6}, \Rightarrow SNR = 26.24$
 $\rightarrow SNR_{dB} = 10 \log_{10}(26.24) = 14.19dB$

Numerical example: Exact BER

i) To compute the BER at $SNR = 20 dB$
 $\rightarrow 10 \log_{10} SNR = 20dB, \Rightarrow SNR = 100$
 $\rightarrow BER = \frac{1}{2} \left(1 - \sqrt{\frac{100}{2+100}}\right) \approx 4.92 \times 10^{-3}$

BER Approximation

i) BER on SNR using **approximation**, (at high SNR)
 $BER = \frac{1}{2} \left(1 - \sqrt{\frac{1}{1+\frac{2}{SNR}}}\right) = \frac{1}{2} \left(1 - (1 + \frac{2}{SNR})^{-\frac{1}{2}}\right)$
 $\approx \frac{1}{2} \left(1 - (1 - \frac{2}{2SNR})\right) = \frac{1}{2} \frac{2}{SNR}$ ■

i) Note that while the BER is **decreasing exponentially** (i.e., $\frac{1}{2} e^{-\frac{1}{2}SNR}$) in a wireline system, it is **decreasing only** by $\frac{1}{2SNR}$ in case of **wireless system**.

Numerical example: Exact and Approx. SNR

i) To compute the SNR required for $BER = 10^{-6}$
 $\rightarrow \frac{1}{2} \left(1 - \sqrt{\frac{SNR}{2+SNR}}\right) = 10^{-6}, \Rightarrow SNR = 4.99 \times 10^5$
 $\rightarrow SNR_{dB} = 10 \log_{10}(4.99 \times 10^5) = 56.98 dB$

ii) Using **approx formulae**, we have $SNR = \frac{1}{2BER}$, further we get $SNR = 57dB$ for $BER = 10^{-6}$

iii) \Rightarrow a wireless channel **requires an extra SNR** of 43 dB (i.e., $10^{4.3}$ times more) to achieve the same BER when compared to AWGN

Numerical example: Exact BER for the given SNR and $L = 2$

i) The average BER can be obtained as,
 $BER = \frac{(1-\lambda)^2}{2} \sum_{l=0}^1 \binom{2+l-1}{l} \left(\frac{1+\lambda}{2}\right)^l$
 $= \left(\frac{1-\lambda}{2}\right)^2 \{2+\lambda\}$

ii) To compute BER when $SNR_{dB} = 20dB$, we have
 $SNR = 10^{\frac{SNR_{dB}}{10}} = 100$ and $\lambda = \sqrt{\frac{100}{2+100}} = 0.9901$
 $BER = 7.2564 \times 10^{-5}$

Numerical example: Approx BER for the given SNR and $L = 2$

i) The approx average BER for $L = 2$ can be obtained as,
 $BER = \binom{3}{2} \left(\frac{1}{2SNR}\right)^2 = \frac{3}{4} \frac{1}{SNR^2}$

ii) To compute approx BER when $SNR_{dB} = 20dB$, we have
 $SNR = 10^{\frac{SNR_{dB}}{10}} = 100$
 $BER = \frac{3}{4} \frac{1}{100^2} = 7.5 \times 10^{-5}$

Numerical example: Approx SNR for the given BER and $L = 2$

i) Given $BER = 10^{-6}$, the SNR can be obtained as,
 $SNR = \sqrt{\frac{3}{4} \frac{1}{10^{-6}}} \Rightarrow SNR = 866.0254 = 29.37dB$

Conclusions

i) **Huge amount of additional power** is required to obtain the desired BER in a single antenna wireless system when compared to AWGN

ii) As the antennas are increased to $L = 2$, the required transmit power P is **decreased** by 27.63 dB

iii) **Increase in antennas** leads to **significant decrease** in BER. This is the advantage of **Diversity**.

Intuition behind the decrease in BER

i) For L rx antennas, we have $BER \approx \binom{2L-1}{L} \left(\frac{1}{2SNR}\right)^L$
for $L = 1$, $BER \approx \frac{1}{2SNR} \propto \frac{1}{SNR}$
for $L = 2$, $BER \approx \frac{3}{4} \frac{1}{SNR^2} \propto \frac{1}{SNR^2}$
for $L = 3$, $BER \approx \frac{5}{4} \frac{1}{SNR^3} \propto \frac{1}{SNR^3}$
in general for L rx antennas, $BER \propto \frac{1}{SNR^L}$

ii) **Deep fade** analysis will help understand this behaviour

Bit-Error Rate, BER

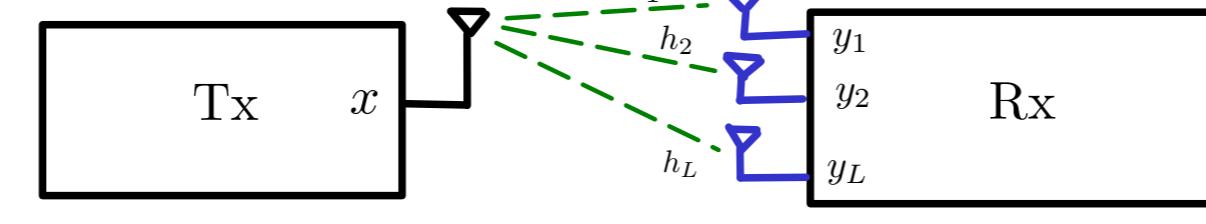
i) BER is a metric to characterize the channel performance
ii) Bits of information undergo errors when transmitted
iii) BER is the average number of bits that are in error
iv) As bits are RV, BER is frequently expressed as
Probability of Bit error P_e , where $0 \leq P_e \leq 0.5$

i) Bits are **modulated** prior to transmission over channel
ii) Binary Phase shift keying (**BPSK**) is one such scheme, where two phases are used $0 \rightarrow x = \sqrt{P}$ and $1 \rightarrow x = -\sqrt{P}$
 \Rightarrow Avg. power of BPSK is P

iii) At the receiver, if
received symbol $y \geq 0$, map $\rightarrow 0$ and
received symbol $y < 0$, map $\rightarrow 1$
(which is also called threshold-based detection)

Multiple receiver antenna wireless communication system

i) Considering wireless system with antenna **diversity** and **optimal MRC combining**



Then the received SNR after **MRC** is given by,
 $SNR_m = \frac{P}{\sigma^2} \{ |h_1|^2 + |h_2|^2 + \dots + |h_L|^2 \} = \frac{P}{\sigma^2} \| \mathbf{h} \|^2 = \frac{P}{\sigma^2} g$,

ii) Assuming $h_i, \forall i$ are IID Rayleigh fading channel coefficients with $E\{|h_i|^2\} = 1, \forall i$, we have
 $\rightarrow g$ is also a **RV** with **Chi-Squared distribution** χ_{2L}^2 with $2L$ degrees of freedom and $F_G(g) = \frac{1}{(L-1)!} g^{L-1} e^{-g}$ ■

iii) We now have $BER = Q(\sqrt{SNR_m}) = Q(\sqrt{g SNR})$ ■

iv) We then have **average BER** in a **Rayleigh fading** channel as
 $BER_{avg} = \int_0^{\infty} Q(\sqrt{g SNR}) F_G(g) dg$
 $= \int_0^{\infty} Q(\sqrt{g SNR}) \frac{1}{(L-1)!} g^{L-1} e^{-g} dg$
 $= \left(\frac{1-\lambda}{2}\right)^L \sum_{l=0}^{L-1} \binom{L+l-1}{l} \left(\frac{1+\lambda}{2}\right)^l$, where
 $\lambda = \sqrt{\frac{SNR}{2+SNR}}$ and $\binom{n}{k} = \frac{n!}{k!(n-k)!}$ ■

v) At **high SNR** using **approx.** we have
 $\frac{1-\lambda}{2} \approx \frac{1}{2SNR}$ and $\frac{1+\lambda}{2} \approx 1$;
 $\Rightarrow BER \approx \left(\frac{1}{2SNR}\right) \sum_{l=0}^{L-1} \binom{L+l-1}{l}$
 $\approx \binom{2L-1}{L} \left(\frac{1}{2SNR}\right)$ ■

BER of different communication systems

i) SNR required to achieve $BER = 10^{-6}$ for same noise power

Channel Type	SNR Required
AWGN	14.19 dB
Rayleigh Fading with $L = 1$	57 dB
Rayleigh Fading with $L = 2$	29.37 dB
Saving in SNR due to diversity	$(57 - 29.37) = 27.63 dB$

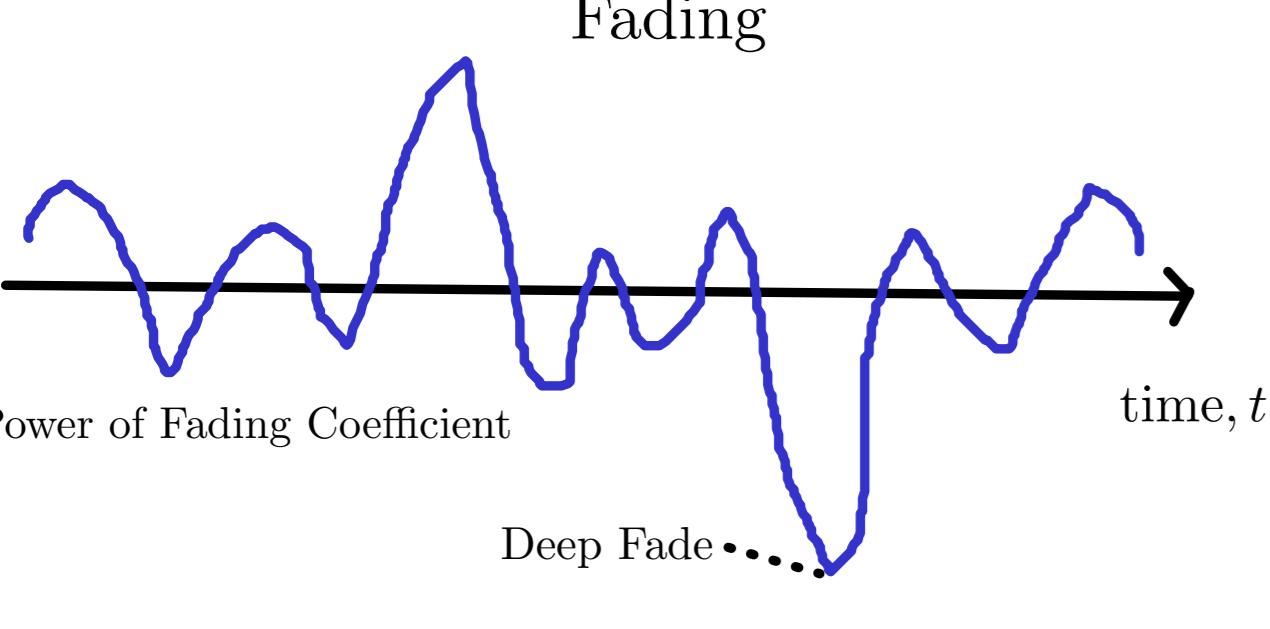
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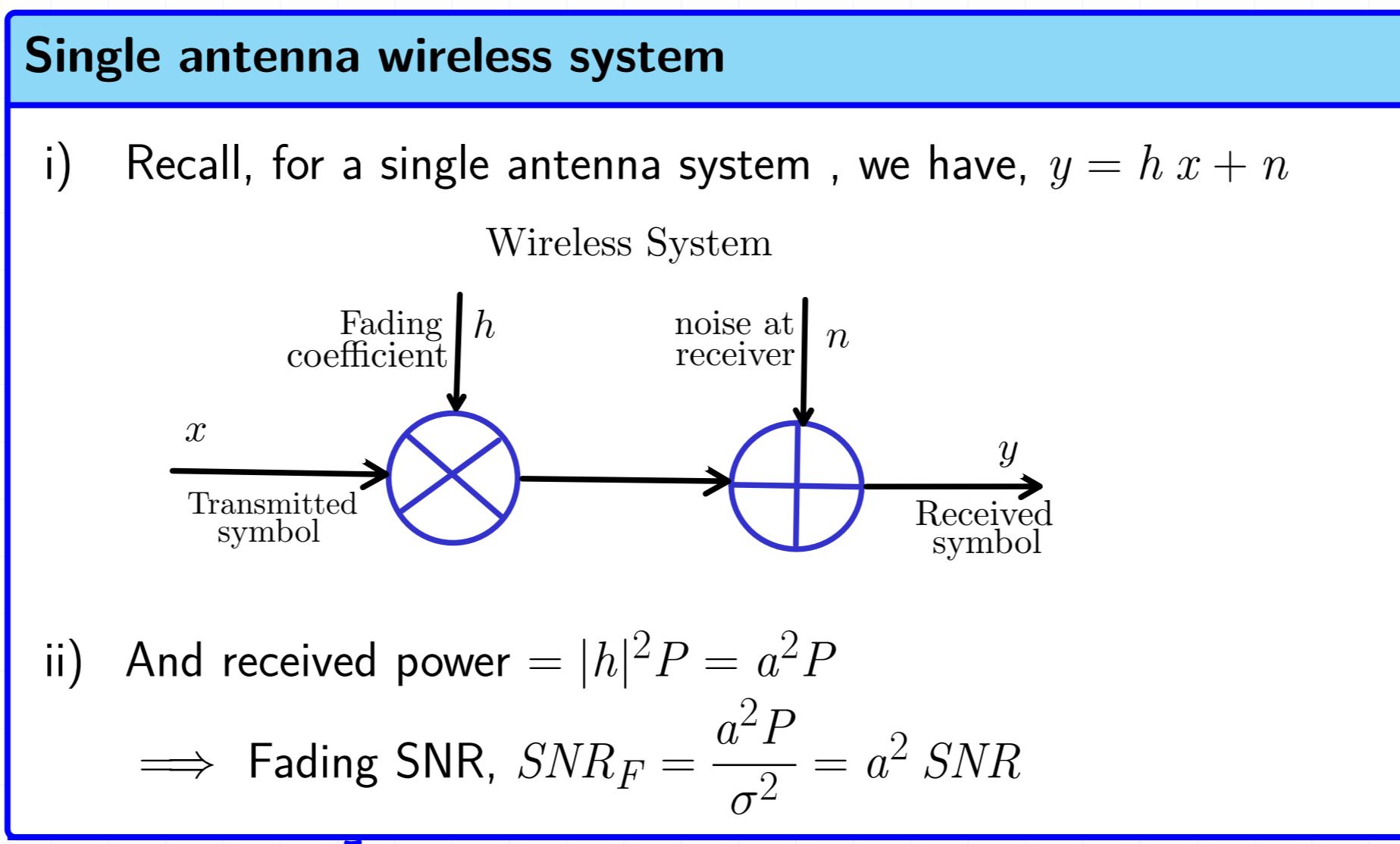
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Performance Analysis (Deep Fade)

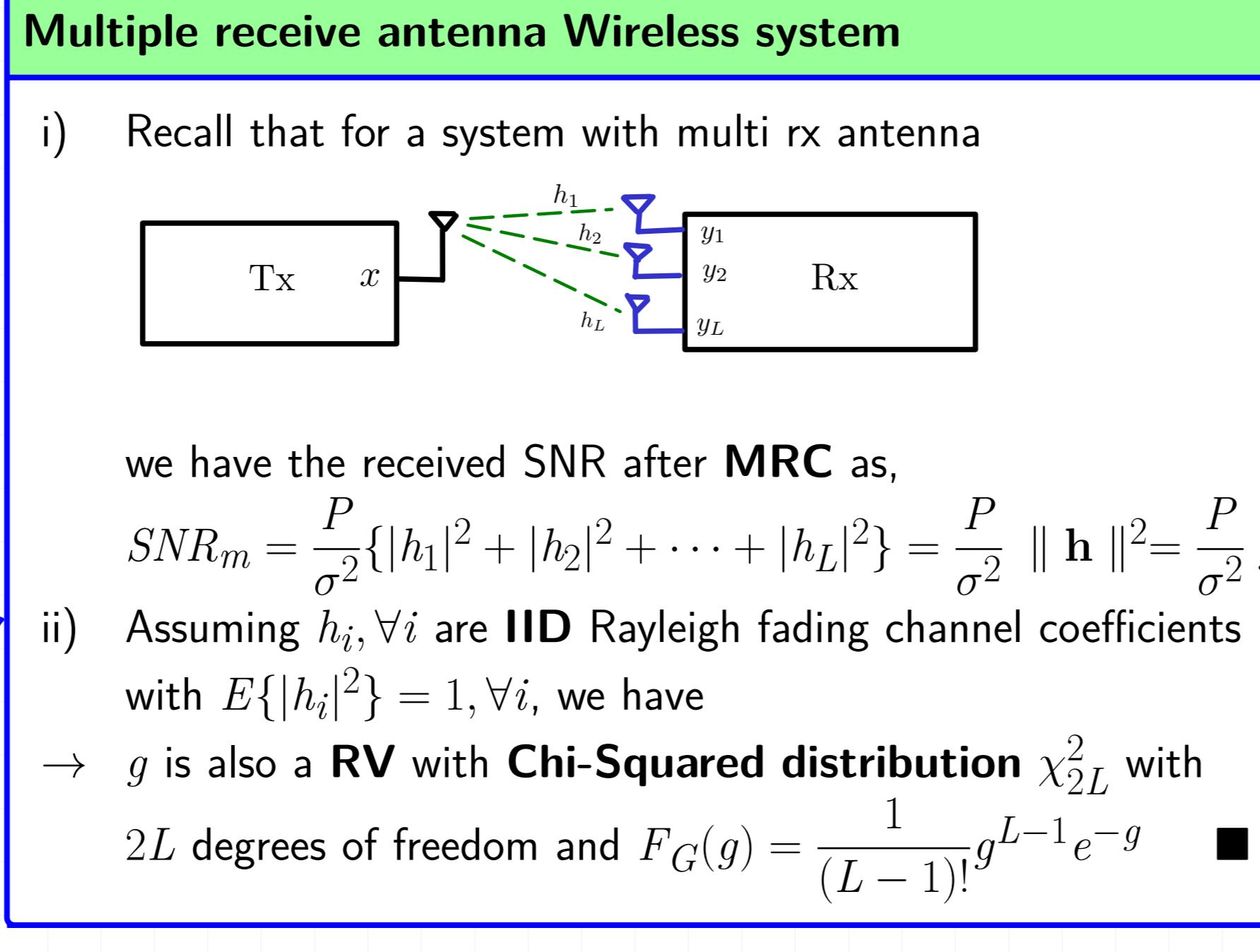
Deep Fade

- i) The power of each channel coefficient $h_i, \forall i$ **varies with time**
- 
- ii) This **Fading** process causes the received power to **vary** and is a key aspect of wireless channels
- iii) The **poor performance** of a wireless comm. system can be explained based on the **Deep Fade** events



Deep Fade Analysis

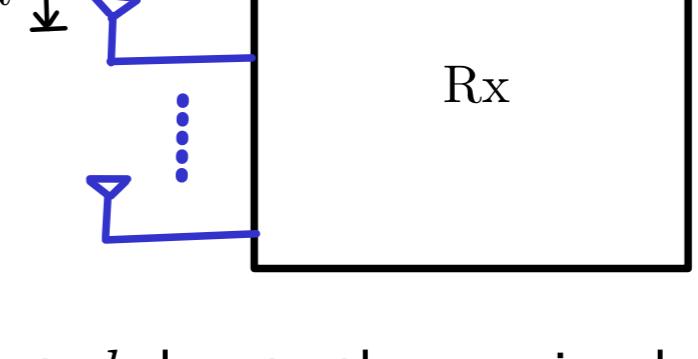
- i) We have **received power** = $a^2 P$ and **noise power** = σ^2
 \rightarrow System is in **deep fade**, if the received signal power is **lower than noise power**
 $\Rightarrow a^2 P < \sigma^2, \Rightarrow a^2 < \frac{\sigma^2}{P} = \frac{1}{SNR}, \Rightarrow a < \sqrt{\frac{1}{SNR}}$
- ii) **Probability of deep fade**, $P_{DF} = Pr(a < \sqrt{\frac{1}{SNR}})$
 $\Rightarrow P_{DF} = \int_0^{\sqrt{\frac{1}{SNR}}} F_A(a) da$
 $= \int_0^{\sqrt{\frac{1}{SNR}}} 2ae^{-a^2} da = 1 - e^{-\frac{1}{SNR}} \blacksquare$
 $\approx 1 - \left(1 - \frac{1}{SNR}\right)$ at high SNR = $\frac{1}{SNR} \blacksquare$
- iii) Note, that **Bit Error Rate** ($\approx \frac{1}{2SNR}$) \propto **Deep Fade**
 \rightarrow The **destructive interference** in wireless channel results in the deep fades
 \rightarrow The performance degrade due to deep fades can be **overcome** by the **principles of diversity**



Deep Fade Analysis with diversity

- i) We have **received power** = $g P$ and **noise power** = σ^2
 \rightarrow System is in **deep fade**, if the received signal power is **lower than noise power**
 $\Rightarrow g P < \sigma^2, \Rightarrow g < \frac{\sigma^2}{P} = \frac{1}{SNR}$
- ii) **Probability of deep fade**, $P_{DF} = Pr(g < \frac{1}{SNR})$
 $\Rightarrow P_{DF} = \int_0^{\frac{1}{SNR}} F_G(g) dg$
 $= \int_0^{\frac{1}{SNR}} \frac{1}{(L-1)!} g^{L-1} e^{-g} dg$
At high SNR, we have $\frac{1}{SNR} \approx 0$,
and note that considered limit is $0 \leq g \leq \frac{1}{SNR}, \Rightarrow e^{-g} \approx 1$
 $\Rightarrow P_{DF} = \int_0^{\frac{1}{SNR}} \frac{1}{(L-1)!} g^{L-1} dg = \frac{1}{(L-1)!} \frac{g^L}{0} \Big|_{0}^{\frac{1}{SNR}}$
 $= \frac{1}{L!} \frac{1}{SNR^L} \propto \frac{1}{SNR^L} \blacksquare$
- iii) With **increasing number of antennas L**, **probability of Deep fade decreases** and this leads to **significant decrease in BER**

Inference

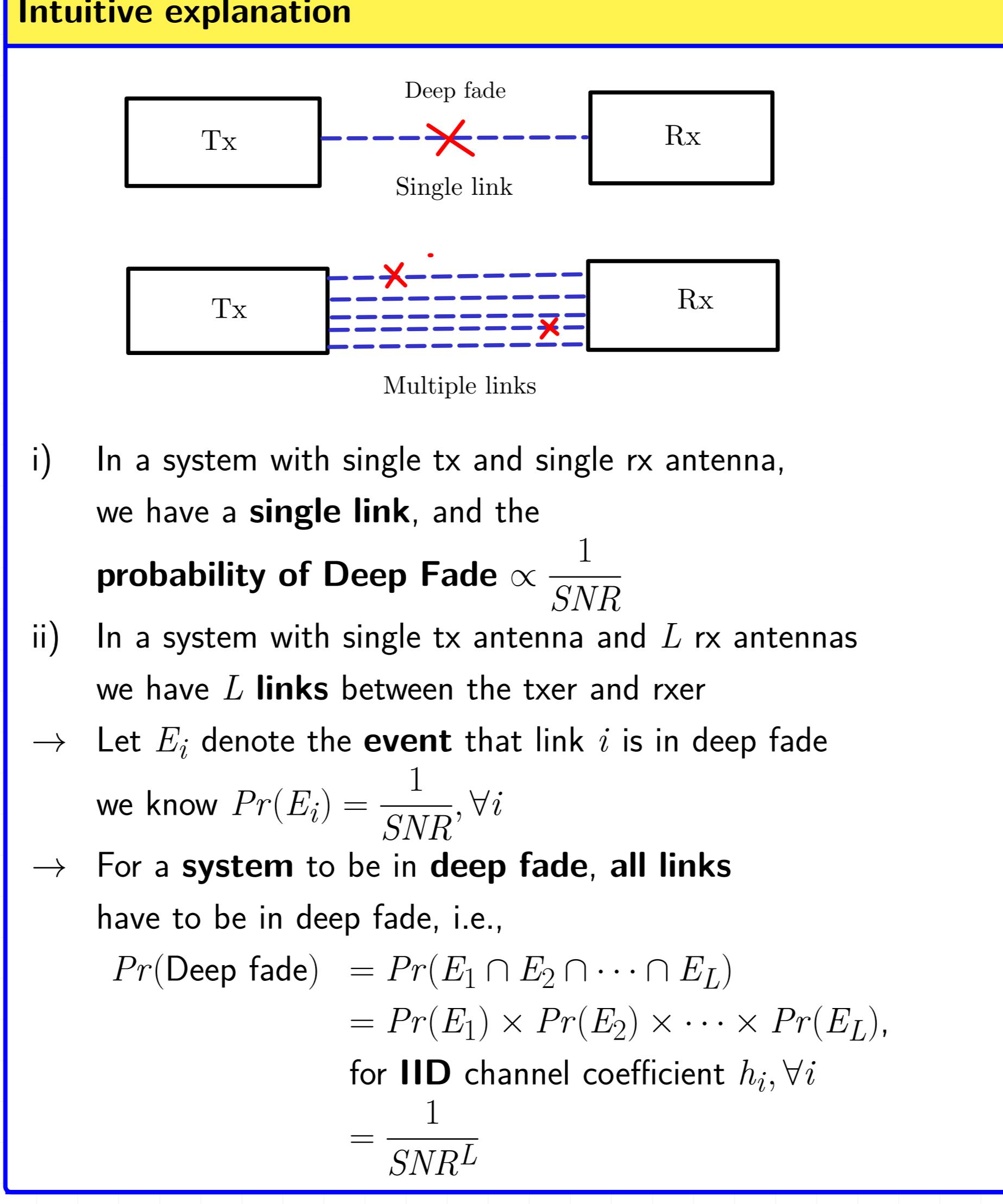
- i) This form of diversity is called **spatial diversity**
- ii) One key assumption is that all the coefficients $h_i, \forall i$ are **Independently** fading
 \rightarrow For this to hold true, antennas have to be placed **sufficiently far apart**
- 
- iii) Let d denote the spacing between the **uniformly spaced** antennas
 \rightarrow For independently fading channels, the condition is given by $d_{min} = \frac{\lambda}{2}$, where d_{min} is the min separation between antennas and $\lambda (= \frac{C}{f_c})$ is the wavelength of the carrier

Numerical example (GSM)

- Q) To compute the d_{min} for a given $f_c = 900MHz$ (GSM)
Sol) We have, $\lambda = \frac{3 \times 10^8 m/s}{900 \times 10^6 / s} = 33.33cm$
 $\therefore d_{min} = 16.66cm$, i.e., distance for independent fading
 \rightarrow Note, that this distance is **not possible** in a regular phone, however, possible in large devices like routers etc.

Numerical example (3G)

- Q) To compute the d_{min} for a given $f_c = 2.3GHz$ (3G)
Sol) We have, $\lambda = \frac{3 \times 10^8 m/s}{2.3 \times 10^9 / s} = 13.04cm$
 $\therefore d_{min} = 6.5cm$, i.e., distance for independent fading
 \rightarrow Note that this distance is within dimensions of a smart phone (3G / 4G)



Principle of Diversity

- i) **Diversity** is used to **combat** the effects of **Deep Fades**
- ii) **Single link system:** Communication fails if the single link is in deep fade
 -
- iii) **Multiple link system:** Communication is **not disrupted** even if two links are deep fade because of many alternative paths
- iv) Ex: **Antenna diversity**
 - Multiple receive antennas introduces **receive diversity**
 - Other forms: Transmit and receive diversity, time diversity, freq. diversity, user diversity etc.

Diversity Order, d

- i) Is a **measure** to **characterize** the diversity of a system

$$d = - \lim_{SNR \rightarrow \infty} \frac{\log P_e(SNR)}{\log SNR},$$
 where $P_e(SNR)$ denotes BER as a function of SNR

Multiple Antenna System

i) **System model:** $y_1 = h_1 x + n_1$
 $y_2 = h_2 x + n_2$, where n_1, n_2 are independent zero-mean Gaussian noise, each with power σ^2 . i.e., $E\{|n_1|^2\} = E\{|n_2|^2\} = \sigma^2$ and $E\{n_1 n_2\} = 0$

ii) **Vector model** (assuming h_1, h_2 as real)

$$\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} h_1 \\ h_2 \end{bmatrix} x + \begin{bmatrix} n_1 \\ n_2 \end{bmatrix}; \quad \mathbf{y} = \mathbf{h} x + \mathbf{n}$$

Receiver

i) **Processing** at the receiver using **weighted combination**

$$\tilde{\mathbf{y}} = w_1 y_1 + w_2 y_2 = \begin{bmatrix} w_1 & w_2 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \mathbf{w}^T \mathbf{y}, \text{ where}$$
 w_1, w_2 are combining weights,
 $\Rightarrow \tilde{\mathbf{y}} = \mathbf{w}^T (\mathbf{h} x + \mathbf{n}) = \mathbf{w}^T \mathbf{h} x + \mathbf{w}^T \mathbf{n}$

ii) **SNR** at the receiver $SNR = \frac{P |\mathbf{w}^T \mathbf{h}|^2}{|\mathbf{w}^T \mathbf{n}|^2}$ where

→ **expected value** of noise power:

$$E\{|\mathbf{w}^T \mathbf{n}|^2\} = E\{(w_1 n_1 + w_2 n_2)^2\}$$
 $= w_1^2 E\{n_1^2\} + w_2^2 E\{n_2^2\} + 2w_1 w_2 E\{n_1 n_2\}$
 $= \sigma^2 (w_1^2 + w_2^2) = \sigma^2 \|\mathbf{w}\|^2$

→ signal power at receiver:
 $\mathbf{w}^T \mathbf{h} = \mathbf{w} \cdot \mathbf{h}$
 $|\mathbf{w} \cdot \mathbf{h}|^2 = \|\mathbf{w}\|^2 \|\mathbf{h}\|^2 \cos^2 \theta$

→ $\therefore SNR = \frac{P \|\mathbf{w}\|^2 \|\mathbf{h}\|^2 \cos^2 \theta}{\sigma^2 \|\mathbf{w}\|^2} = \frac{P}{\sigma^2} \|\mathbf{h}\|^2 \cos^2 \theta$

Optimal Receiver (MRC)

i) **SNR is maximum** when $\theta = 0$,
i.e., \mathbf{w} is in the same direction of \mathbf{h} (or $\mathbf{w} \propto \mathbf{h}$)

$$\Rightarrow SNR_{max} = \frac{P(h_1^2 + h_2^2)}{\sigma^2}, \text{ when } \mathbf{w} = \frac{1}{\sqrt{h_1^2 + h_2^2}} \begin{bmatrix} h_1 \\ h_2 \end{bmatrix}$$

→ When channel coefficients are complex, we use

$$\tilde{\mathbf{y}} = \mathbf{w}^H \mathbf{y} = \begin{bmatrix} w_1^* & w_2^* \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} \text{ and } \mathbf{w} = \frac{1}{\sqrt{|h_1|^2 + |h_2|^2}} \begin{bmatrix} h_1 \\ h_2 \end{bmatrix}$$

→ This receiver is called **Maximal Ratio Combiner (MRC)**

Analytical Example with $L = 1$

Q) Compute the diversity order of a wireless system with $L = 1$ antenna

Sol) In this case, we have $P_e(SNR) \approx \frac{1}{2SNR}$ under high SNR

$$d = - \lim_{SNR \rightarrow \infty} \frac{\log \frac{1}{2SNR}}{\log SNR} = \lim_{SNR \rightarrow \infty} 1 + \frac{\log 2}{\log SNR} = 1$$

Analytical Example with multiple antennas

Q) Compute the diversity order of a wireless system with L antennas

Sol) In this case, we have

$$d = - \lim_{SNR \rightarrow \infty} \frac{\log \left(\frac{2L-1}{L} \right) \left(\frac{1}{2SNR} \right)^L}{\log SNR},$$
 $= - \lim_{SNR \rightarrow \infty} \frac{-L \log SNR + \log \left(\frac{2L-1}{L} \right) \frac{1}{2L}}{\log SNR}$
 $= \lim_{SNR \rightarrow \infty} L - \frac{\log \left(\frac{2L-1}{L} \right) \frac{1}{2L}}{\log SNR} = L$

Analytical Example for wireline system

Q) Compute the diversity order of a wireline system

Sol) In this case, we have

$$d = - \lim_{SNR \rightarrow \infty} \frac{\log \frac{1}{2} e^{-\frac{1}{2} SNR}}{\log SNR} = - \lim_{SNR \rightarrow \infty} \frac{\log \frac{1}{2} - \frac{1}{2} SNR}{\log SNR},$$
 $= \lim_{SNR \rightarrow \infty} \frac{\frac{1}{2} SNR}{\log SNR} = \infty, \text{ using L'Hospital's rule}$

\Rightarrow A AWGN channel is a combination of ∞ independently fading links

Example of a MRC

i) Given, $h_1 = \frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}} j$ and $h_2 = \frac{1}{\sqrt{2}} - \frac{1}{\sqrt{2}} j$ and noise power $\sigma^2 = \frac{1}{2} = -3dB$

Sol) We have $\begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}} j \\ \frac{1}{\sqrt{2}} - \frac{1}{\sqrt{2}} j \end{bmatrix} x + \begin{bmatrix} n_1 \\ n_2 \end{bmatrix}$

→ then, optimal MRC vector $\mathbf{w} = \frac{1}{\sqrt{2}} \begin{bmatrix} \frac{1}{\sqrt{2}} + \frac{1}{\sqrt{2}} j \\ \frac{1}{\sqrt{2}} - \frac{1}{\sqrt{2}} j \end{bmatrix}$

→ Further we obtain, $|\mathbf{w}^T \mathbf{h}|^2 = 2$ and $\|\mathbf{w}\|^2 = 1$,

→ $\therefore SNR = \frac{P |\mathbf{w}^T \mathbf{h}|^2}{|\mathbf{w}^T \mathbf{n}|^2} = \frac{2P}{\frac{1}{2}} = 4P$

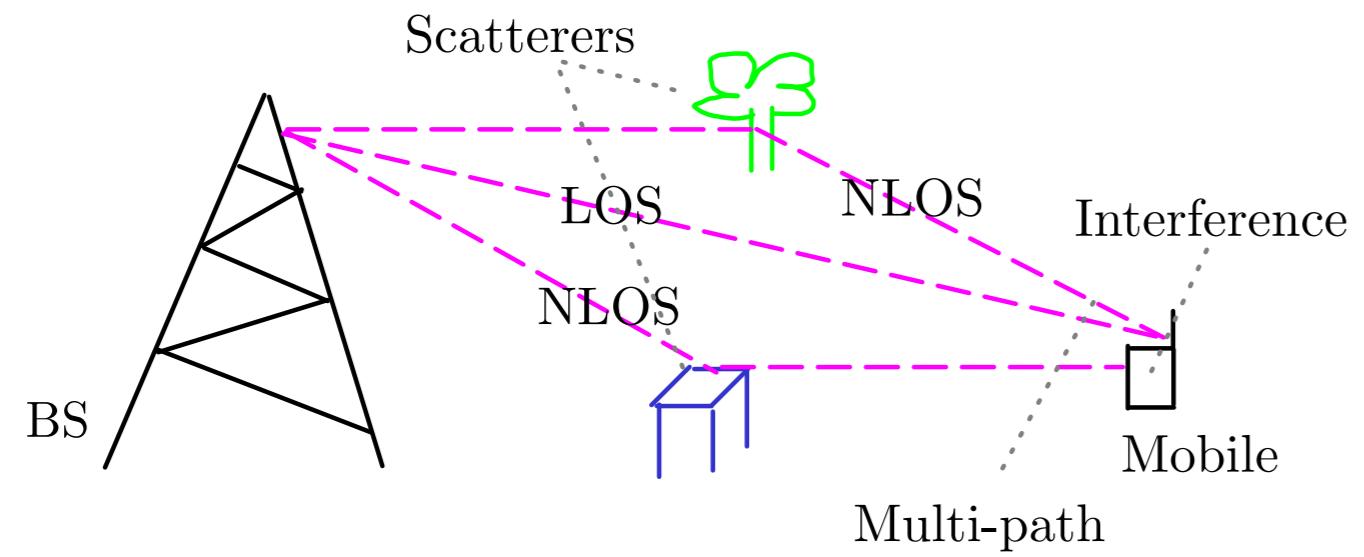
Generalizing the MRC for L receiver antennas

i) Given $E\{|n_i|^2\} = \sigma^2, \forall i$
 $E\{n_i n_j\} = 0, \forall i \neq j$, $\begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_L \end{bmatrix} = \begin{bmatrix} h_1 \\ h_2 \\ \vdots \\ h_L \end{bmatrix} x + \begin{bmatrix} n_1 \\ n_2 \\ \vdots \\ n_L \end{bmatrix}$

and $\tilde{\mathbf{y}} = w_1^* y_1 + w_2^* y_2 + \dots + w_L^* y_L$
 $\Rightarrow \tilde{\mathbf{y}} = \mathbf{w}^H (\mathbf{h} x + \mathbf{n}) = \mathbf{w}^H \mathbf{h} x + \mathbf{w}^H \mathbf{n}$

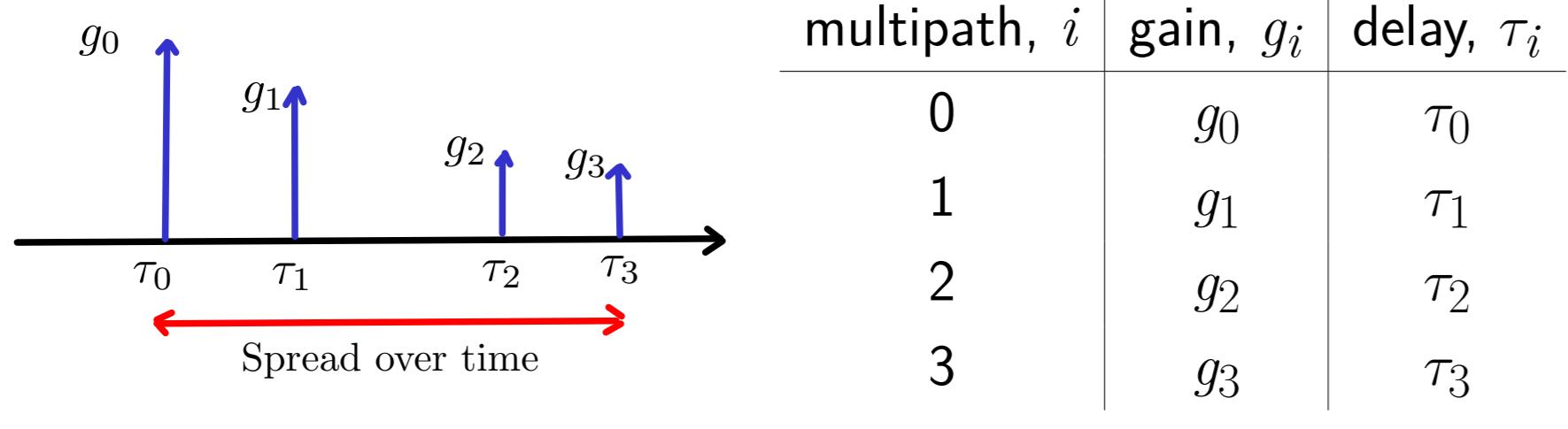
ii) For maximum SNR, we choose MRC with $\mathbf{w} = \frac{\mathbf{h}}{\|\mathbf{h}\|}$
 $\therefore SNR = \frac{\|\mathbf{h}\|^2 P}{\sigma^2} = \frac{(|h_1|^2 + |h_2|^2 + \dots + |h_L|^2) P}{\sigma^2}$

Power Profile



- i) **Channel response** is $h(t) = \sum_{i=0}^{L-1} a_i \delta(t - \tau_i)$, where a_i is attenuation and τ_i is the delay of i^{th} multipath
- ii) Then **Power profile** of the i^{th} multipath is $\phi(t) = \sum_{i=0}^{L-1} |a_i|^2 \delta(t - \tau_i) = \sum_{i=0}^{L-1} g_i \delta(t - \tau_i)$, where $g_i = |a_i|^2$ is gain of i^{th} path

Example: Consider an $L = 4$ multipath channel



- i) The spread of delay over which different signal components are arriving is called **Delay spread**
- ii) **Maximum Delay spread** σ_{τ}^{\max} characterizes the delay spread, $\sigma_{\tau}^{\max} = \tau_{L-1} - \tau_0$, where τ_0 is the delay of the first arriving component and τ_{L-1} is the delay of the last arriving component
- iii) Ex: For $L = 4$, let $\tau_0 = 0\mu s$ and $\tau_3 = 5\mu s$, then $\sigma_{\tau}^{\max} = 5\mu s$

RMS Delay Spread

i) Practically, there might be large number of spurious paths with negligible power below the noise floor
In such case, max. delay spread is not an **appropriate** metric

ii) **RMS Delay spread** (root mean square) is considered in such scenario. We now consider
 $b_i = \frac{g_i}{\sum_{j=0}^{L-1} g_j}$, the fraction of power in i^{th} path

Average delay by weighing each delay by fraction of power
 $\bar{\tau} = \sum_{i=0}^{L-1} b_i \tau_i = \frac{\sum_{i=0}^{L-1} g_i \tau_i}{\sum_{i=0}^{L-1} g_i}$

Root Mean square delay spread

$$\sigma_{\tau} = \sqrt{b_0(\tau_0 - \bar{\tau})^2 + \dots + b_{L-1}(\tau_{L-1} - \bar{\tau})^2} = \sqrt{\frac{\sum_{i=0}^{L-1} g_i (\tau_i - \bar{\tau})^2}{\sum_{i=0}^{L-1} g_i}}$$

Impact of channel power profile

i) Consider a transmitted signal with symbols $\{S_0, S_1, S_2, \dots\}$

where T is the symbol time

ii) Further consider 2-path wireless channel (i.e., $L = 2$) where

$$h(t) = a_0 \delta(t - \tau_0) + a_1 \delta(t - \tau_1) = \delta(t - \tau_0) + \delta(t - \tau_1)$$
, where $a_0 = a_1 = 1$

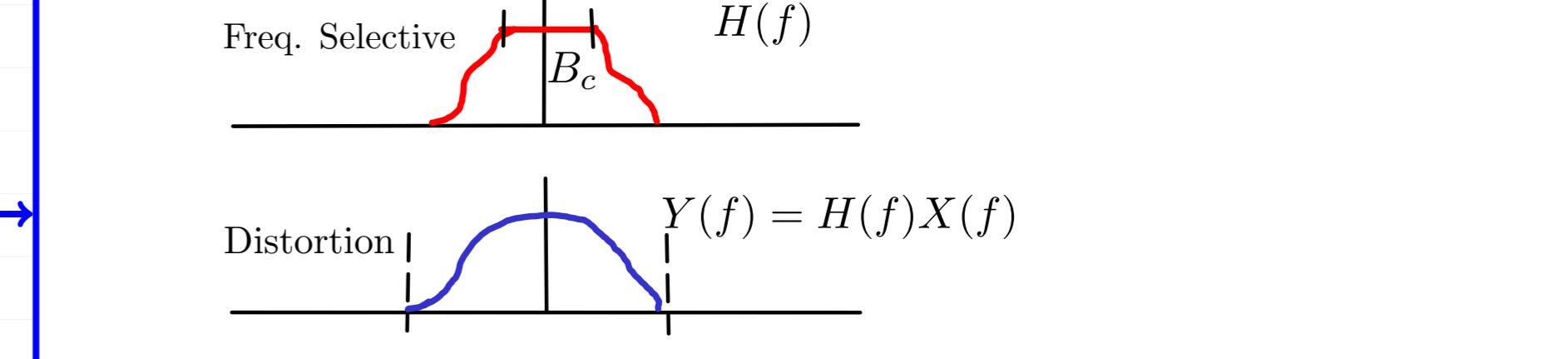
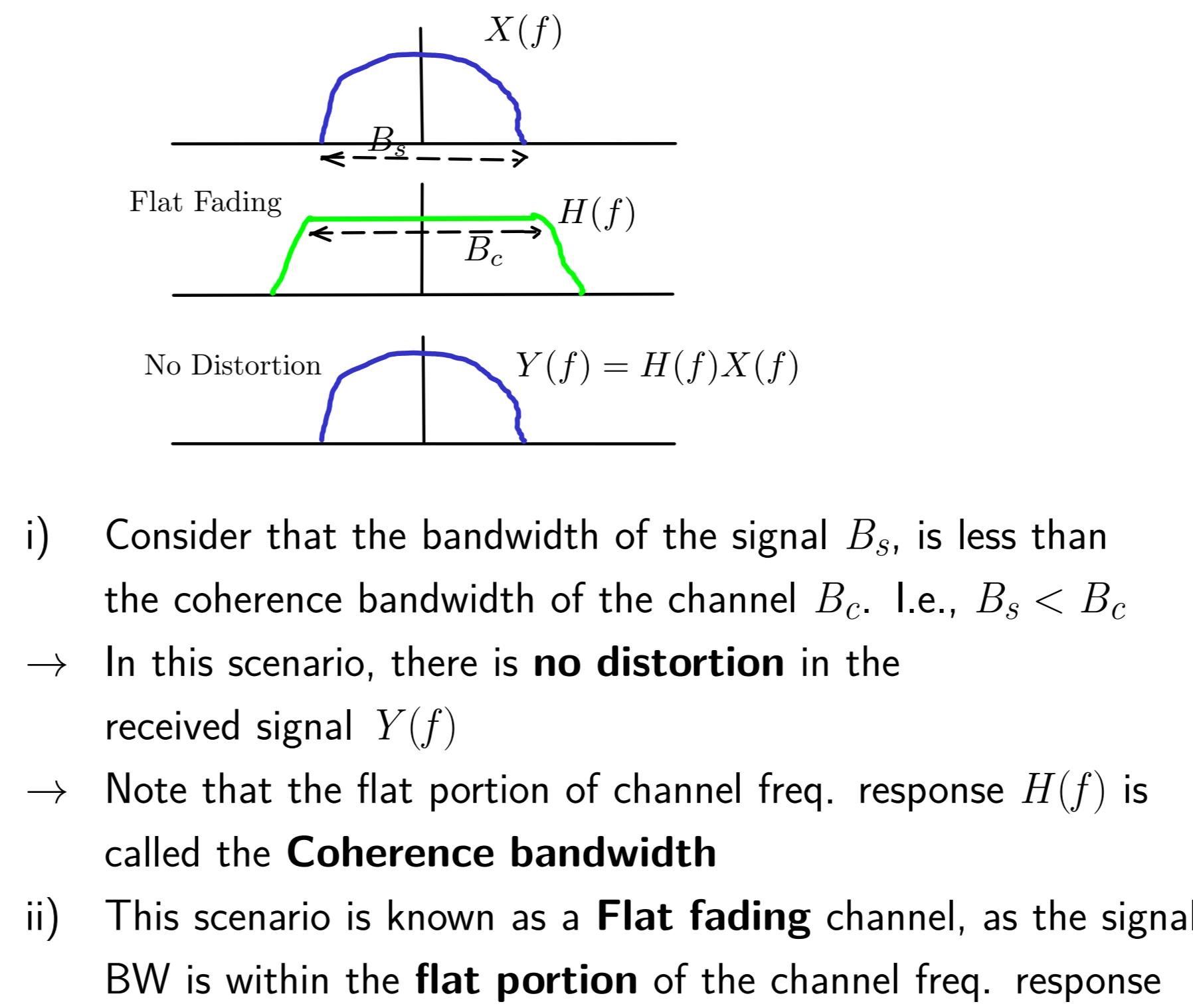
Example:

- Q) Compute RMS delay, given $L = 4$, $\tau_i = \{0, 1, 3, 5\} \mu s \forall i$, $g_i = \{0.01, 0.1, 1, 0.1\} = \{-20, -10, 0, -10\} \text{dB } \forall i$, Given,
-
- Multipath power profile
- $$\bar{\tau} = \frac{0.01 \times 0 + 0.1 \times 1 + 1 \times 3 + 0.1 \times 5}{0.01 + 0.1 + 1 + 0.1} = 2.9752 \mu s$$
- $$\rightarrow \text{RMS Delay } \sigma_{\tau} = \sqrt{\frac{0.01 \times (0 - 2.9752)^2 + \dots + 0.1 \times (5 - 2.9752)^2}{0.01 + 0.1 + 1 + 0.1}} = 0.8573 \mu s$$
- $$\rightarrow \tau_{\max} = 5 - 0 = 5 \mu s$$

Inter Symbol Interference (ISI)

- i) The two delayed signals arriving at the receiver add-up,
-
- Signal delayed by τ_0
- Signal delayed by τ_1
- Because of the difference in delays the adjacent symbols interfere
- ii) Inter symbol interference (ISI) occurs when $(\tau_1 - \tau_0) > T$ i.e., **delay spread** is greater than **symbol time** or $\sigma_{\tau} > T$
- ISI is undesirable since it leads to distortion of the original signal
- To avoid ISI, we need the symbol time T to be larger than delay spread i.e., $T > \sigma_{\tau}$
- iii) **Typical outdoor** delay spread in a wireless channels is $2\mu s$ (difference in LOS and NLOS path in a cell is $1KM$,)

Signal Bandwidth $B_s < \text{Channel Coherence Bandwidth } B_c$



Typical Component

- i) Focusing on a typical channel component $e^{-j2\pi f \tau_i}$, At $f = 0$ the phase = 0 and at $f = \frac{1}{2\tau_i}$ the phase = π
→ Note that the phase (freq. response) has changed significantly within $f = \frac{1}{2\tau_i}$, i.e., within a bandwidth = $2 \times \frac{1}{2\tau_i} = \frac{1}{\tau_i}$
∴ In general, approximately the BW of the channel = $\frac{1}{\sigma_{\tau}} = B_c$
- ii) → The **larger** the **delay spread**, the **smaller** is the **coherence BW**
- iii) Typically $\sigma_{\tau} = 2\mu s$, $\Rightarrow B_c = 500KHz$

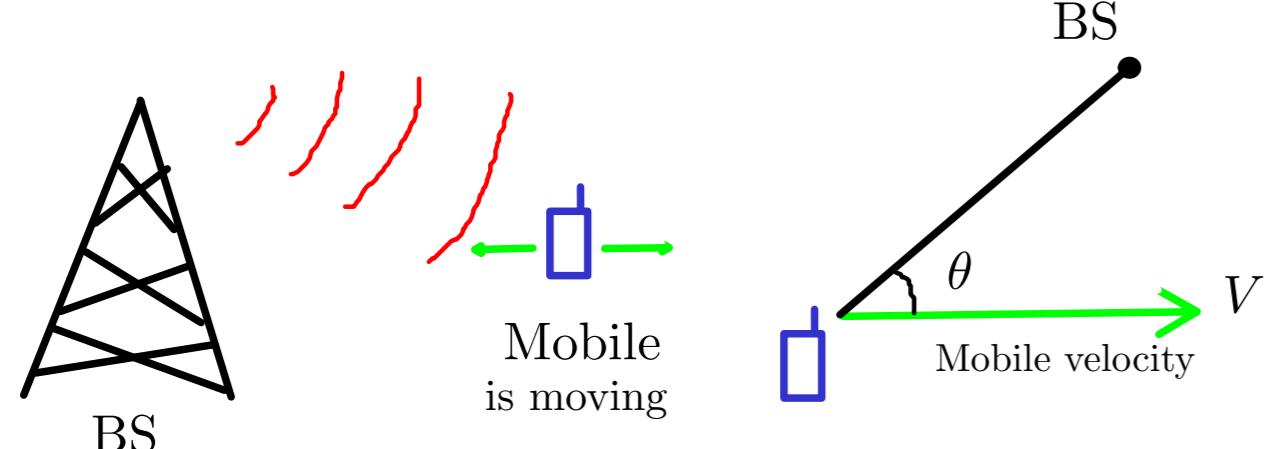
Interpretation from freq. and time domain analysis

Freq. domain	Time domain	Implication
If $B_s > B_c$	$\Rightarrow \sigma_{\tau} > T$	Freq. Selective Fading and ISI
If $B_s < B_c$	$\Rightarrow \sigma_{\tau} < T$	Freq. Flat Fading and no ISI

Example:

- i) Considering $f_d = 143Hz$, the coherence time $T_c = \frac{1}{4f_d} = \frac{1}{4 \times 143} = 1.7ms$
- The channel is approx. constant for (or changing after) $1.7ms$

Doppler Effect



- i) In a scenario where the mobile is **moving**, will causes the freq. of the received signal to **change**
→ This is known as the **Doppler effect**
→ It is the change in freq. of the received signal due to motion between Txer and the Rxer
- ii) Assuming the mobile is moving at an angle θ with the BS, we have the doppler freq. $f_d = \frac{V \cos \theta}{C} \times f_c$, where ■ V is velocity of mobile, C is speed of light and f_c is carrier freq.
- iii) Frequency of rxed signal $f_r = (1 + \frac{V \cos \theta}{C}) f_c$. If mobile is
→ moving towards BS (i.e., $0 < \theta \leq \frac{\pi}{2}$), then $f_r \geq f_c$
→ moving away from BS (i.e., $\frac{\pi}{2} < \theta \leq \pi$), then $f_r \leq f_c$

Impact of Doppler on Wireless channel

i) Recall fading coefficient is $h = \sum_{i=0}^{L-1} a_i e^{-j2\pi f_c \tau_i}$

ii) When the mobile is moving, the delay τ_i changes with time
 $\Rightarrow \tau_i(t) = \tau_i \pm \frac{V \cos \theta t}{C}$ and

$$h = \sum_{i=0}^{L-1} a_i e^{-j2\pi f_c (\tau_i \pm \frac{V \cos \theta t}{C})} e^{\mp j2\pi f_d t} = h(t)$$

Note, the channel coefficient is time-varying and hence known as time selective channel

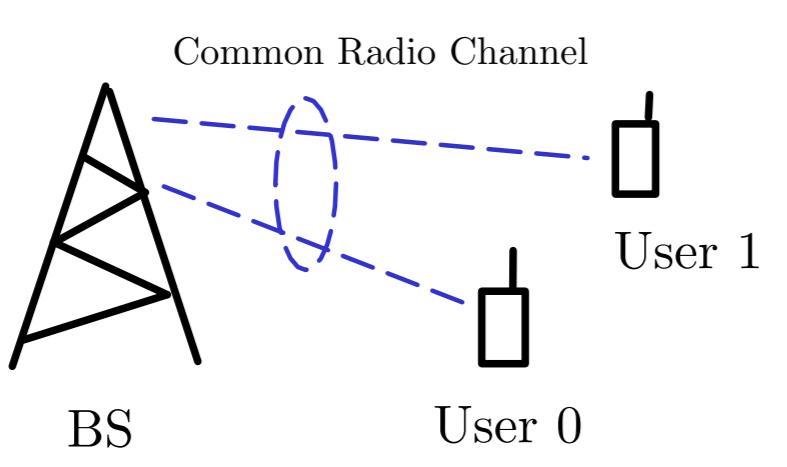
iii) Mobility → Doppler → Time-varying phase → Time selective

iv) Considering the phase factor
at $t = 0$ the phase = 0, and at $t = \frac{1}{4f_d}$ the phase = $\frac{\pi}{2}$ has changed significantly
→ **Coherence time** $T_c (= \frac{1}{4f_d})$ is defined as the time over which channel is approx. constant ■

Code Division Multiple Access

Code Division Multiple Access

- i) CDMA is a key technology used for multiple access in various 3G standards like WCDMA, HSDPA / HSUPA, CDMA 2000 and 1×EVDO
- ii) Unlike wireline where each user is allocated a dedicated channel, in multiple access, users share a common radio channel
- iii) TDMA and FDMA are other examples of multiple access technologies
 - TDMA: In time division multiple access, different users are allocated different time slots
 - FDMA: In frequency division multiple access, different users are allocated different bands. FDMA was used in 1G comm. systems
- iv) Similarly, in CDMA different users are allocated different codes
 - CDMA uses the concept of codes for multiple access



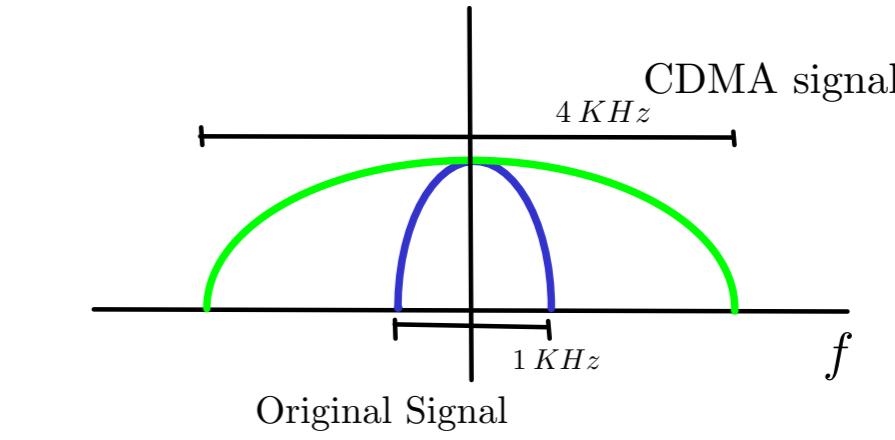
Example: 2-user multiple access system

- i) Consider codes C_0 and C_1 , given by
 $C_0 = \begin{bmatrix} 1 & 1 & 1 & 1 \end{bmatrix}$ and $C_1 = \begin{bmatrix} 1 & -1 & -1 & 1 \end{bmatrix}$
 and transmitted symbol of user-0 and user-1 as a_0 and a_1
- ii) Before transmission each symbol is multiplied with resp. code as
 $a_0 C_0 = \begin{bmatrix} a_0 & a_0 & a_0 & a_0 \end{bmatrix}$,
 $a_1 C_1 = \begin{bmatrix} a_1 & -a_1 & -a_1 & a_1 \end{bmatrix}$
 and the combined sum is transmitted
 $a_0 C_0 + a_1 C_1 = \begin{bmatrix} a_0 + a_1 & a_0 - a_1 & a_0 - a_1 & a_0 + a_1 \end{bmatrix}$
 As a result 4 chips are transmitted in a symbol duration.
- iii) To extract the individual symbols at the receiver, each user correlates the received symbol with its code i.e.,
 user-0: $1(a_0 + a_1) + 1(a_0 - a_1) + 1(a_0 - a_1) + 1(a_0 + a_1) = 4a_0$
 user-1: $1(a_0 + a_1) - 1(a_0 - a_1) - 1(a_0 - a_1) + 1(a_0 + a_1) = 4a_1$
 There by multiple users are able to simultaneously receive signals on a common channel

Chip Time and Bandwidth expansion in CDMA

- i) Considering an example with symbol rate of 1 kbps , i.e., 1000 sym/sec , we have time per symbol $T_s = \frac{1}{1000} \text{ sec}$. i.e., Bandwidth required $BW = \frac{1}{T_s} = 1 \text{ KHz}$.

- ii) However in CDMA (with 4 chips/sym),
 $\text{chip time } T_c = \frac{T_s}{N} = \frac{1}{4} \text{ ms} = 0.25 \text{ ms}$
 $\Rightarrow \text{new bandwidth} = \frac{1}{T_c} = 4 \text{ KHz}$
 Original bandwidth has spread in CDMA.



- Spectrum of the original signal is spread by a factor of N , where N is the length of the code. As a result, it is also called spread spectrum system
- CDMA codes are also known as the spreading codes. And N is termed as spreading factor.

Properties of the CDMA PN sequence

i) Balance Property:

- The number of (-1)'s (=8) \approx the number of (1)'s (=7)

ii) Run length Property:

- A Run is defined as the string of continuous values
 This property states, that $\frac{1}{2^m}$ of total runs are of length n
 There are a total of 8 runs in the above example and
 1 run of $n = 4$ (Note # of state variables = 4)
 $1 = (\frac{1}{2^3} \times 8)$ run of $n = 3$, $2 = (\frac{1}{2^2} \times 8)$ run of $n = 2$
 $4 = (\frac{1}{2^1} \times 8)$ run of $n = 1$

ii) Auto Correlation Property:

- Auto correlation $r(d) \triangleq \frac{1}{N} \sum_{n=0}^N C(n)C(n-d)$
 Note that, $r(d=0) = \frac{1}{N} \sum_{n=0}^N C(n)^2 = \frac{N}{N} = 1$
 However, $r(d=2) = \frac{1}{N} \sum_{n=0}^N C(n)C(n-2) = \frac{-1}{N} = \frac{-1}{15}$
 $C(n) = [-1 -1 -1 -1 1 1 1 1 -1 1 1 -1 1 -1 1]$
 $C(n-2) = [-1 1 -1 -1 -1 1 1 1 -1 1 1 -1 1 -1 1]$
 $C(n) \times C(n-2) = [1 -1 1 1 -1 -1 1 1 -1 -1 1 1 -1 1 1]$
 $\Rightarrow r(d) = \begin{cases} 1 & \text{if } d = 0, \\ \frac{-1}{N} \rightarrow 0 & \text{if } d \neq 0 \text{ (and large } N\text{).} \end{cases}$

BER of a single-user CDMA system

- i) Using $SNR = SNR_{BPSK} = \frac{P}{\sigma^2}$, we have

$$SNR_{CDMA} = N|h|^2 SNR$$

$$\Rightarrow BER = \frac{1}{2} \left(1 - \sqrt{\frac{N \text{ SNR}}{2 + N \text{ SNR}}} \right) \text{ (Exact)} \blacksquare$$

$$\approx \frac{1}{2N \text{ SNR}} \text{ (Approx. for high SNR)} \blacksquare$$

- Q) Consider CDMA system with code length $N = 256$ and $SNR = \frac{P}{\sigma^2} = 15 \text{ dB}$, compute BER over Rayleigh fading channel

$$\text{Sol. } BER = \frac{1}{2} \left(1 - \sqrt{\frac{256 \times 10^{1.5}}{2 + 256 \times 10^{1.5}}} \right) = 6.175 \times 10^{-5}$$

CDMA system with Frequency selective channel (or ISI)

- i) Consider an L -tap channel with $\{h(0), h(1), \dots, h(L-1)\}$ and the transmitted symbols as $\{x(0), \dots, x(N)\}$. Then,

$$y(n) = h(0)x(n) + h(1)x(n-1) + \dots + w(n) = \sum_{l=0}^{L-1} h(l)x(n-l) + w(n)$$

i.e., a clear case of inter symbol interference

- ii) Considering a single-user CDMA system $x(n) = a_0 C_0(n)$, then

$$y(n) = \sum_{l=0}^{L-1} h(l)a_0 C_0(n-l) + w(n)$$

- iii) Performing correlation operation at the receiver,
 $r(0) = \frac{1}{N} \sum_{n=0}^{N-1} y(n)C_0(n)$
 $= \frac{1}{N} \sum_n \sum_m h(m)a_0 C_0(n)C_0(m)$
 $= \sum_l h(l)a_0 \frac{1}{N} \sum_n C_0(n-l)C_0(n) + \tilde{w}$, we have

$$\text{autocorrelation, } \frac{1}{N} \sum_n C_0(n-l)C_0(n) = \begin{cases} 1 & \text{if } l = 0, \\ \frac{-1}{N} \approx 0 & \text{if } l \neq 0 \end{cases}$$

$$\Rightarrow r(0) = h(0)a_0 + \tilde{w} \blacksquare$$

- iv) Similarly, performing correlation with $C_0(n-1)$, we get
 $r(1) = h(1)a_0 + \tilde{w} \blacksquare$

- v) Vector model can be written as, $\mathbf{r} = \mathbf{h}a_0 + \tilde{\mathbf{w}}$,

$$\begin{bmatrix} r(0) \\ \vdots \\ r(L-1) \end{bmatrix} = \begin{bmatrix} h(0) \\ \vdots \\ h(L-1) \end{bmatrix} a_0 + \begin{bmatrix} \tilde{w}(0) \\ \vdots \\ \tilde{w}(L-1) \end{bmatrix}$$

Rake Receiver: Carrying out Optimal processing using MRC on \mathbf{r} , i.e., $\frac{\mathbf{h}^H}{\|\mathbf{h}\|^2} \mathbf{r}$, we obtain

$$i) SNR_{rake} = \frac{\|\mathbf{h}\|^2 P}{\sigma^2/N} = N \|\mathbf{h}\|^2 SNR \blacksquare$$

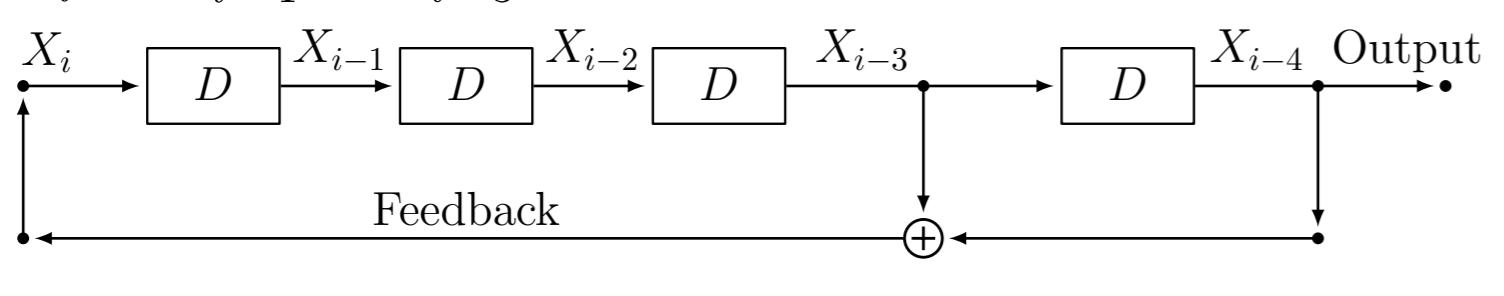
$\|\mathbf{h}\|^2$ is also known as multi-path diversity in CDMA

- ii) As a result $BER = \binom{2L-1}{L} \left(\frac{1}{2NSNR} \right)^L$, where L is the # of taps of freq. selective channel and coefficients $h(0) \dots h(L-1)$ are IID rayleigh with unity avg power

Example: Given $L = 3$ taps for FSC, $N = 256$ and $SNR = 15 \text{ dB}$. Then $BER = \binom{5}{3} \left(\frac{1}{2 \times 256 \times 10^{1.5}} \right)^3 = 2.36 \times 10^{-12}$

Code generation of CDMA

- i) The code is typically a Pseudo Noise (PN) sequence (random). A PN seq. is generated by a linear feedback shift register (LFSR)
- ii) A LFSR with D states can go through $2^D - 1$ states (except all zeros $[0 \dots 0]$)
- A LFSR which goes through $2^D - 1$ states is called maximal length LFSR
- iii) Example: Consider a LFSR with 4 state variables and $X_i = X_{i-4} \oplus X_{i-3}$.



This has $15 (= 2^4 - 1)$ states, as a result is a maximal length LFSR

- The code is obtained by mapping the output X_{i-4} as, $0 \rightarrow 1$ and $1 \rightarrow -1$

Example: LFSR for $N = 4$

State Variables \rightarrow	X_{i-1}	X_{i-2}	X_{i-3}	X_{i-4}	$X_i (= X_{i-4} \oplus X_{i-3})$
States ↓ 1	1	1	1	1	0
2	0	1	1	1	0
3	0	0	1	1	0
4	0	0	0	1	1
5	1	0	0	0	0
6	0	1	0	0	0
7	0	0	1	0	1
8	1	0	0	1	1
9	1	1	0	0	0
10	0	1	1	0	1
11	1	0	1	1	0
12	0	1	0	1	1
13	1	0	1	0	1
14	1	1	0	1	1
15	1	1	1	0	1
All 1's 16	1	1	1	1	0

The CDMA PN sequence for this case is,

$$C(n) = [-1 -1 -1 -1 1 1 1 1 -1 1 1 -1 1 -1 1]$$

System model of a CDMA (single-user) system

- i) Consider a single-user CDMA system, i.e., User-0 transmits symbol a_0 , uses $C_0(n) \in \{-1, +1\} \forall n$ with length N , and let Transmitted symbol $x(n) = a_0 C_0(n)$
 Received symbol $y(n) = hx(n) + w(n)$, where h is Rayleigh fading coeff
 $w(n)$ is IID Gaussian noise $\mathcal{N}(0, \sigma^2)$
 $\Rightarrow E\{w(n)\} = 0 \quad \forall n$ and
 $E\{w(n)w^*(m)\} = \begin{cases} \sigma^2 & \text{if } n = m, \\ 0 & \text{if } n \neq m \end{cases}$
- ii) At receiver, $y(n)$ is correlated with $C_0(n)$
 $r_{y0} = \frac{1}{N} \sum_{n=0}^N y(n)C_0(n) = \frac{1}{N} \sum_{n=0}^N [hx(n) + w(n)]C_0(n)$
 $= \frac{1}{N} \sum_{n=0}^N ha_0 C_0^2(n) + \frac{1}{N} \sum_{n=0}^N w(n)C_0(n)$
 $= ha_0 + \tilde{w}$, where $\tilde{w} = \frac{1}{N} \sum_{n=0}^N w(n)C_0(n)$

SNR of a single-user CDMA system

Further, analyzing the noise component

$$\begin{aligned} E\{\tilde{w}\} &= \frac{1}{N} \sum_{n=0}^N E\{w(n)\} C_0(n) = 0 \\ E\{\tilde{w}\tilde{w}^*\} &= E\left\{ \left(\frac{1}{N} \sum_n w(n) C_0(n) \right) \left(\frac{1}{N} \sum_m w^*(m) C_0^*(m) \right) \right\} \\ &= \frac{1}{N^2} \sum_n \sum_m E\{w(n)w^*(m)\} E\{C_0(n)C_0^*(m)\} \\ &= \frac{1}{N^2} \sum_n E\{C_0(n)^2\} E\{C_0(n)^2\} = \frac{1}{N} \end{aligned} \blacksquare$$

Note that, Noise power decreased by N

$$iii) SNR_{CDMA} = \frac{\text{Signal Power}}{\text{Noise Power}} = \frac{E\{|h|^2|a_0|^2\}}{\sigma^2/N} = N|h|^2 P \blacksquare$$

This important factor N is called spreading gain or processing gain of CDMA system

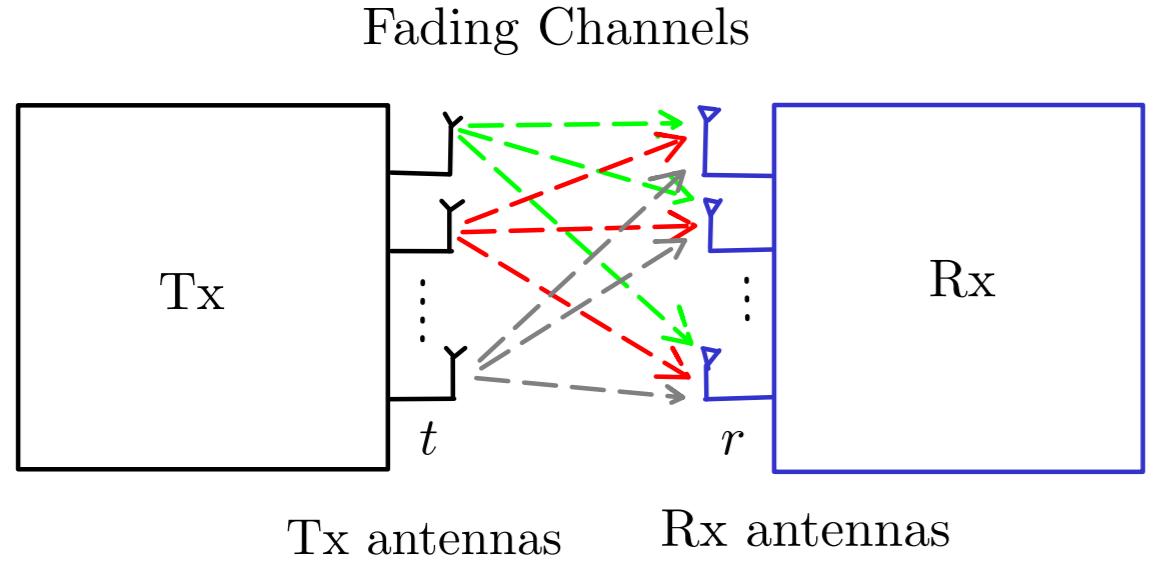
System model of CDMA (Multi-user) system

- i) Consider a system with {user-0, user-1} transmitting symbols $\{a_0, a_1\}$, using $\{C_0(n), C_1(n)\}$ with length N , and let Transmitted symbol $x(n) = a_0 C_0(n) + a_1 C_1(n)$
 Received symbol $y(n) = hx(n) + w(n)$
 $= ha_0 C_0(n) + ha_1 C_1(n) + w(n)$
- ii) At receiver, $y(n)$ is correlated with $C_0(n)$
 $r_{y0} = \frac{1}{N} \sum_{n=0}^N y(n)C_0(n)$
 $= \frac{1}{N} \sum_{n=0}^N [ha_0 C_0(n) + ha_1 C_1(n) + w(n)]C_0(n)$
 $= ha_0 + ha_1 \frac{1}{N} \sum_{n=0}^N C_0(n)C_1(n) + \tilde{w}$
 Note, the multi-user interference (MUI) component above $\left[= ha_1 \frac{1}{N} \sum_{n=0}^N C_0(n)C_1$

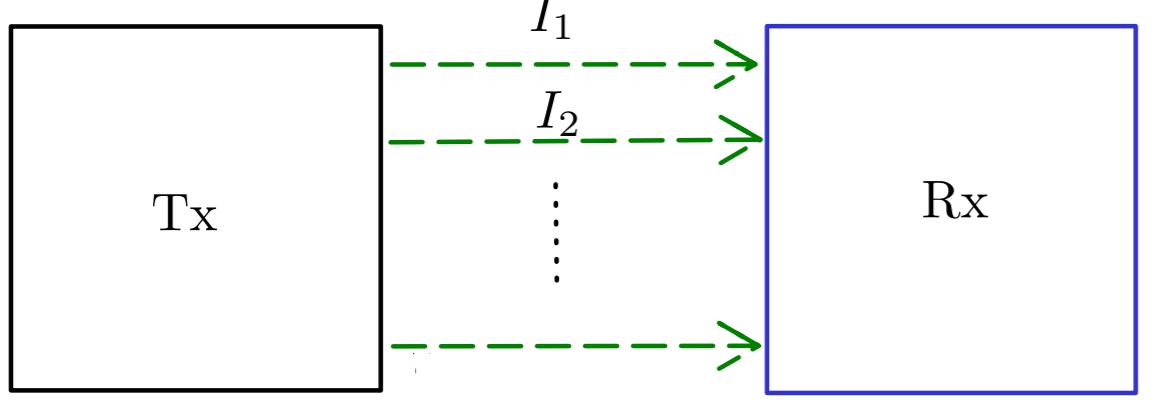
Multiple Input Multiple Output

Multiple Input Multiple Output

- i) This is a key technology for 3G/4G wireless systems
- Multiple Inputs \Rightarrow Multiple Transmit Antennas (t)
- Multiple Outputs \Rightarrow Multiple Receive Antennas (r)
- MIMO \Rightarrow System with $r \times t$ antennas



- ii) MIMO leads to
 - \rightarrow Increase in Data Rates by transmitting several information streams in parallel also termed **Spatial Multiplexing**.



- \rightarrow Increase in Reliability due to **increased diversity**

MIMO receiver (when $r = t$)

- i) Objective is to recover the transmit vector \mathbf{x} at the receiver from the receive vector, $\mathbf{y} = \mathbf{Hx} + \mathbf{w}$
- ii) In a simple scenario, where $r = t$, $\mathbf{H}_{r \times r}$ is a square matrix (i.e., # of Rx antennas = # of Tx antennas).
- \rightarrow Then, if \mathbf{H} is invertible, we get $\hat{\mathbf{x}} = \mathbf{H}^{-1}\mathbf{y}$ using approx. that $\mathbf{y} \approx \mathbf{Hx}$, i.e., system of linear equations

MIMO

MISO: Transmit Beamforming ($r = 1$)

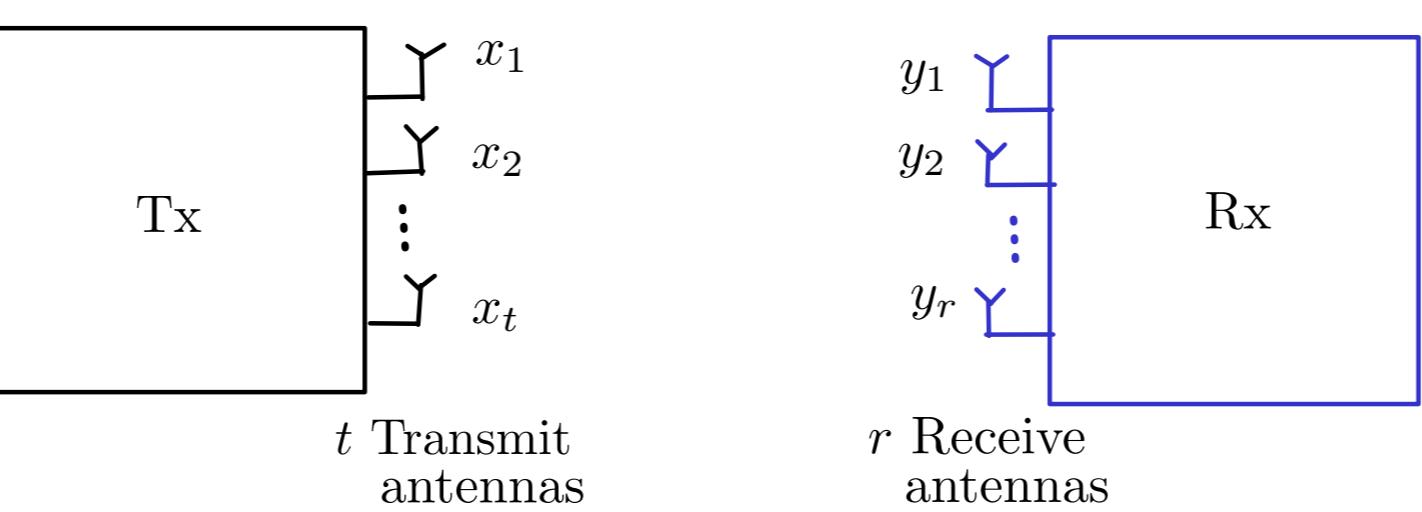
- i) Considering a 2×1 MISO system, we have
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- $$\mathbf{y} = \begin{bmatrix} h_1 & h_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \mathbf{w}, \text{ then What is the optimal tx scheme } \mathbf{x}?$$
- ii) Transmit Beamformer uses, $\mathbf{x} = \frac{1}{\|\mathbf{h}\|} \begin{bmatrix} h_1^* \\ h_2^* \end{bmatrix} x$, where x is the transmit symbol and the unit vector $\frac{1}{\|\mathbf{h}\|} \begin{bmatrix} h_1^* \\ h_2^* \end{bmatrix}$ is the transmit beamformer
 - iii) At the receiver, we then have
- $$\mathbf{y} = \begin{bmatrix} h_1 & h_2 \end{bmatrix} \begin{bmatrix} h_1^* \\ h_2^* \end{bmatrix} \frac{1}{\|\mathbf{h}\|} x + \mathbf{w} = \|\mathbf{h}\| x + \mathbf{w}$$

ALAMOUTI Space Time Block Code

- i) Does NOT require knowledge of CSI at Transmitter. Very useful from practical viewpoint
- ii) Consider a 2×1 MISO system transmit symbols x_1 and x_2 using transmit vectors $\begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$ and $\begin{bmatrix} -x_2^* \\ x_1^* \end{bmatrix}$ at time instants 1 and 2 respectively.
- iii) The, the symbol y that we receive at time instant 1 and 2 is $y_1 = \begin{bmatrix} h_1 & h_2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + w_1$ and $y_2 = \begin{bmatrix} h_1 & h_2 \end{bmatrix} \begin{bmatrix} -x_2^* \\ x_1^* \end{bmatrix} + w_2$

System Model

- i) **Transmitter:** t symbols, $\mathbf{x}^T = [x_1, x_2, \dots, x_t]$ can be transmitted on the t transmit antennas. Where \mathbf{x} is the t dimensional transmit vector.



- ii) **Receiver:** r samples, $\mathbf{y}^T = [y_1, y_2, \dots, y_r]$ are received across r receive antennas. \mathbf{y} is the r dimensional receive vector.

- iii) Channel is represented by matrix $\mathbf{H}_{r \times t}$, which is defined by

$$\mathbf{y}_{r \times 1} = \mathbf{H}_{r \times t} \mathbf{x}_{t \times 1} + \mathbf{w}_{r \times 1}$$

$$\mathbf{H} \triangleq \begin{bmatrix} h_{11} & h_{12} & \dots & h_{1t} \\ h_{21} & h_{22} & \dots & h_{2t} \\ \vdots & & & \\ h_{r1} & h_{r2} & \dots & h_{rt} \end{bmatrix}_{r \times t}, \quad \mathbf{x}_{t \times 1} \xrightarrow{\text{MIMO Channel}} \mathbf{y}_{r \times 1}$$

where h_{ij} is the channel coefficient between i^{th} receive and j^{th} transmit antenna

- iv) In general, in a $r \times t$ system we have $y_l = h_{l1}x_1 + h_{l2}x_2 + \dots + h_{lt}x_t + w_l$, where y_l is the sample received at l^{th} antenna.

Example: 3×2 MIMO system

3 receive antennas and 2 transmit antennas

$$\begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}_{3 \times 1} = \begin{bmatrix} h_{11} & h_{12} \\ h_{21} & h_{22} \\ h_{31} & h_{32} \end{bmatrix}_{3 \times 2} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}_{2 \times 1} + \begin{bmatrix} w_1 \\ w_2 \\ w_3 \end{bmatrix}_{3 \times 1}$$

$$y_1 = h_{11}x_1 + h_{12}x_2 + w_1$$

$$\Rightarrow y_2 = h_{21}x_1 + h_{22}x_2 + w_2$$

$$y_3 = h_{31}x_1 + h_{32}x_2 + w_3$$

Least Square problem (when $r > t$)

- i) However, if $r > t$ (i.e., # of Rx antennas $>$ # of Tx antennas), then \mathbf{H} is not a square matrix, \Rightarrow inverse of \mathbf{H} does not exist.
 - $\rightarrow r > t \Rightarrow$ # of Equations $>$ # of unknowns
 - ii) Alternatively, define $\mathbf{e} \triangleq \mathbf{y} - \mathbf{Hx}$, now find \mathbf{x} that minimizes the error, i.e.,
- $$\min \| \mathbf{e} \|^2 = \min \| \mathbf{y} - \mathbf{Hx} \|^2$$
- iii) To find the $\hat{\mathbf{x}}$, differentiate the RHS w.r.t vector \mathbf{x} and set equal to zero. Further we use the below two properties,
 - a) $\frac{\partial(\mathbf{c}^T \mathbf{x})}{\partial \mathbf{x}} = \frac{\partial(\mathbf{x}^T \mathbf{c})}{\partial \mathbf{x}} = \mathbf{c}$ and
 - b) $\frac{\partial(\mathbf{x}^T \mathbf{P} \mathbf{x})}{\partial \mathbf{x}} = 2\mathbf{P}\mathbf{x}$, where $\mathbf{P} = \mathbf{P}^T$

BER for BPSK with ZF receiver

- i) Considering transmit symbol power $E\{|x_i|^2\} = P$ and each h_{ij} as IID complex Gaussian with unity avg. power,
- $$BER = \binom{2L-1}{L} \left(\frac{1}{2SNR} \right)^L, \text{ where } L = r - t + 1$$
- i.e., equivalent to MRC system with $r - t + 1$ receive antennas

- Q.) Given, 3×2 MIMO system, $SNR = 25dB$ and ZF receiver, compute the BER.

$$\text{Sol)} \text{ Avg BER} = \binom{3}{2} \left(\frac{1}{2 \times 10^{2.5}} \right)^2 = 7.5 \times 10^{-6}$$

SNR due to Transmit Beamforming

- i) We then have $SNR = \frac{\|\mathbf{h}\|^2 E\{|x|^2\}}{\sigma^2} = \frac{\|\mathbf{h}\|^2 P}{\sigma^2}$
- \Rightarrow Trasmit Beamforming is able to achieve SAME SNR as the MRC
- ii) However, major CHALLENGE is that Channel State Information (CSI) is to be known as the transmitter

- Q.) Given a 1×2 MISO system, with CSI $h_1 = 2 + j$ and $h_2 = 1 - 2j$ available at the transmitter, compute the optimal beamforming vector.

$$\text{Sol)} \text{ We have } \mathbf{x} = \frac{1}{\|\mathbf{h}\|} \begin{bmatrix} h_1^* \\ h_2^* \end{bmatrix} x = \frac{1}{\sqrt{10}} \begin{bmatrix} 2-j \\ 1+2j \end{bmatrix} x \text{ and}$$

$$\mathbf{y} = \begin{bmatrix} 2+j & 1-2j \end{bmatrix} \begin{bmatrix} 2-j \\ 1+2j \end{bmatrix} \frac{1}{\sqrt{10}} x + \mathbf{w} = \sqrt{10}x + \mathbf{w}$$

Alamouti Receiver (Orthogonal STBC)

- ii) By rearranging y_1 and y_2^* and process them jointly, we have
- $$\begin{bmatrix} y_1 \\ y_2^* \end{bmatrix} = \begin{bmatrix} h_1 & h_2 \\ h_2^* & -h_1^* \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} w_1 \\ w_2^* \end{bmatrix} = \begin{bmatrix} h_1 & h_2 \\ h_2^* & -h_1^* \end{bmatrix} x_1 + \begin{bmatrix} h_2 \\ -h_1^* \end{bmatrix} x_2 + \begin{bmatrix} w_1 \\ w_2^* \end{bmatrix}$$
- $$\Rightarrow \mathbf{y} = \mathbf{c}_1 x_1 + \mathbf{c}_2 x_2 + \mathbf{w}, \text{ where } \mathbf{c}_1 \triangleq \begin{bmatrix} h_1 \\ h_2^* \end{bmatrix} \text{ and } \mathbf{c}_2 \triangleq \begin{bmatrix} h_2 \\ -h_1^* \end{bmatrix}$$

Note that \mathbf{c}_1 and \mathbf{c}_2 are ORTHOGONAL!. i.e., $\mathbf{c}_1^H \mathbf{c}_2 = 0$

- iii) To recover x_1 from \mathbf{y} , we process as
- $$\mathbf{c}_1^H \mathbf{y} = \frac{\mathbf{c}_1^H}{\|\mathbf{c}_1\|} \mathbf{c}_1 x_1 + \frac{\mathbf{c}_1^H}{\|\mathbf{c}_1\|} \mathbf{c}_2 x_2 + \frac{\mathbf{c}_1^H}{\|\mathbf{c}_1\|} \mathbf{w} = \|\mathbf{c}_1\| x_1 + 0 + \tilde{\mathbf{w}} \text{ where,}$$
- $$E\{\tilde{\mathbf{w}}\} = E\left\{ \frac{\mathbf{c}_1^H}{\|\mathbf{c}_1\|} \mathbf{w} \mathbf{w}^H \frac{\mathbf{c}_1}{\|\mathbf{c}_1\|} \right\} = \frac{\mathbf{c}_1^H}{\|\mathbf{c}_1\|} E\{\mathbf{w} \mathbf{w}^H\} \frac{\mathbf{c}_1}{\|\mathbf{c}_1\|} = \sigma^2$$
- iv) Then, $SNR = \frac{\|\mathbf{c}_1\|^2 E\{|x_1|^2\}}{\sigma^2} = \frac{\|\mathbf{h}\|^2 \frac{P}{2}}{\sigma^2} = \frac{1}{2} SNR$ of MRC!

Special Cases

SIMO: Single Input Multiple Output system ,

$$\text{i.e., } t = 1 \text{ and } \begin{bmatrix} y_1 \\ \vdots \\ y_r \end{bmatrix}_{r \times 1} = \begin{bmatrix} h_1 \\ \vdots \\ h_r \end{bmatrix}_{r \times 1} x + \begin{bmatrix} w_1 \\ \vdots \\ w_r \end{bmatrix}_{r \times 1}$$

MISO: Multiple Input Single Output system ,

$$\text{i.e., } r = 1 \text{ and } y = [h_1 \dots h_t]_{1 \times t} \begin{bmatrix} x \\ \vdots \\ x \end{bmatrix}_{t \times 1} + w$$

SISO: Single Input Single Output system ,

$$\text{i.e., } r = t = 1 \text{ and } y = hx + w$$

Noise Vector analysis

- i) Each w_i is Gaussian with zero-mean and σ^2 variance , $E\{w_i\} = 0$ and $E\{|w_i|^2\} = E\{w_i w_i^*\} = \sigma^2$

- ii) Further, w_i are IID, i.e.,

$$E\{w_i w_j^*\} = \begin{cases} \sigma^2 & \text{if } i = j, \\ E\{w_i\} E\{w_j^*\} = 0 & \text{if } i \neq j \end{cases}$$

\rightarrow As a result noise Covariance matrix \mathbf{R} can be obtained as

$$E\{\mathbf{w} \mathbf{w}^H\} = E\left\{ \begin{bmatrix} w_1 \\ \vdots \\ w_r \end{bmatrix} \begin{bmatrix} w_1^* & \dots & w_r^* \end{bmatrix} \right\} = \begin{bmatrix} \sigma^2 & 0 & \dots & 0 \\ 0 & \sigma^2 & & 0 \\ \vdots & & \ddots & \\ 0 & 0 & \dots & \sigma^2 \end{bmatrix}$$

$$\mathbf{R} = \sigma^2 \mathbf{I}_{r \times r}, \text{ where } \mathbf{I} \text{ is an identity matrix}$$

Least Square Cost Function

$$\begin{aligned} f(\mathbf{x}) &\triangleq \|\mathbf{y} - \mathbf{Hx}\|^2 = (\mathbf{y} - \mathbf{Hx})^T (\mathbf{y} - \mathbf{Hx}) \\ &= (\mathbf{y}^T - \mathbf{x}^T \mathbf{H}^T)(\mathbf{y} - \mathbf{Hx}) \\ &= \mathbf{y}^T \mathbf{y} - \mathbf{x}^T \mathbf{H}^T \mathbf{y} - \mathbf{y}^T \mathbf{Hx} + \mathbf{x}^T \mathbf{H}^T \mathbf{Hx} \\ &= \mathbf{y}^T \mathbf{y} - 2\mathbf{x}^T \mathbf{H}^T \mathbf{y} + \mathbf{x}^T \mathbf{H}^T \mathbf{Hx} \end{aligned}$$

Differentiating $f(\mathbf{x})$ and equating to 0, we have

$$\begin{aligned} \frac{\partial f(\mathbf{x})}{\partial \mathbf{x}} &= 0 - 2\mathbf{H}^T \mathbf{y} + 2\mathbf{H}^T \mathbf{Hx} = 0 \\ \mathbf{H}^T \mathbf{Hx} &= \mathbf{H}^T \mathbf{y} \\ \hat{\mathbf{x}} &= (\mathbf{H}^T \mathbf{H})^{-1} \mathbf{H}^T \mathbf{y} \quad \blacksquare \end{aligned}$$

Zero Forcing Receiver

- i) $\hat{\mathbf{x}} = (\mathbf{H}^T \mathbf{H})^{-1} \mathbf{H}^T \mathbf{y}$, is called as the Zero Forcing receiver
- \rightarrow Generalizing it to the case where \mathbf{H} is complex, we have $\hat{\mathbf{x}} = (\mathbf{H}^H \mathbf{H})^{-1} \mathbf{H}^H \mathbf{y}$ \blacksquare

- ii) Note that $\mathbf{H}^H \mathbf{H}$ is a $t \times t$ square matrix and $(\mathbf{H}^H \mathbf{H})^{-1} \mathbf{H}^H$ is the Pseudo (also called left) inverse of \mathbf{H} i.e., $(\mathbf{H}^H \mathbf{H})^{-1} (\mathbf{H}^H \mathbf{H}) = \mathbf{I}$

Example: ZF receiver for 3×2 MIMO system

- i) Consider a MIMO system with $r = 3$ and $t = 2$

$$\begin{bmatrix$$

Singular Value Decomposition (SVD)

- i) The channel matrix $\mathbf{H}_{r \times t}$ is SV decomposed for $r \geq t$ when

$$\mathbf{H}_{r \times t} = [\mathbf{U}_{r \times t} \ \Sigma_{t \times t} \ \mathbf{V}_{t \times t}^H]$$

$$\Sigma_{t \times t} = \begin{bmatrix} \sigma_1 & 0 & \cdots & 0 \\ 0 & \sigma_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & 0 \\ 0 & 0 & \cdots & \sigma_t \end{bmatrix}_{t \times t}$$

- ii) Columns of \mathbf{U} are Orthonormal

i.e., $\mathbf{u}_i^H \mathbf{u}_j = \begin{cases} \|\mathbf{u}_i\|^2 = 1 & \text{if } i = j, \\ 0 & \text{if } i \neq j \end{cases}$

$$\mathbf{U}^H \mathbf{U} = \begin{bmatrix} \mathbf{u}_1^H \\ \mathbf{u}_2^H \\ \vdots \\ \mathbf{u}_t^H \end{bmatrix} \begin{bmatrix} \mathbf{u}_1 & \mathbf{u}_2 & \cdots & \mathbf{u}_t \end{bmatrix} = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & 0 \\ 0 & 0 & \cdots & 1 \end{bmatrix} = \mathbf{I}_{t \times t}$$

→ Similarly $\mathbf{V}^H \mathbf{V} = \mathbf{V} \mathbf{V}^H = \mathbf{I}_{t \times t}$, i.e. \mathbf{V} is UNITARY matrix

- iii) $\Sigma_{t \times t}$ is a diagonal matrix with $\sigma_1 \geq \sigma_2 \geq \cdots \geq \sigma_t \geq 0$,
 → Singular values are non-negative and are arranged in decreasing order
 → The # of non-zero singular values indicate the rank of the channel matrix

MIMO-II

Capacity of a MIMO wireless system

- i) Considering spatial multiplexing in MIMO i.e., $\tilde{y}_i = \sigma_i \tilde{x}_i + \tilde{w}_i$ where σ_i is gain of i^{th} channel, $E\{|\tilde{x}_i|^2\} = P_i$ is power of i^{th} transmitted symbol and \tilde{w}_i is Gaussian with var σ^2 , we have $SNR_i = \frac{\sigma_i^2 P_i}{\sigma^2}$

- ii) Then we have Shannon rate = $\log_2 \left(1 + \frac{\sigma_i^2 P_i}{\sigma^2} \right)$
 The sum rate = $\sum_{i=0}^t \log_2 (1 + SNR_i)$
 iii) Considering transmit power as P , the capacity of MIMO

$$C = \max_P \sum_{i=0}^t \log_2 \left(1 + \frac{\sigma_i^2 P_i}{\sigma^2} \right)$$

sub.to $\sum_{i=0}^t P_i = P$

which is a constrained optimization problem

SVD in MIMO Systems

- i) SVD can be employed in MIMO comm. systems as a tool. Consider the MIMO model

$$\mathbf{y}_{r \times 1} = \mathbf{H}_{r \times t} \mathbf{x}_{t \times 1} + \mathbf{w}_{r \times 1}, \text{ now substituting } \mathbf{H} = \mathbf{U} \Sigma \mathbf{V}^H$$

$$\mathbf{y} = \mathbf{U} \Sigma \mathbf{V}^H \mathbf{x} + \mathbf{w}$$

- ii) At the receiver, multiply the received vector with \mathbf{U}^H , we have

$$\mathbf{U}^H \mathbf{y} = \mathbf{U}^H \mathbf{U} \Sigma \mathbf{V}^H \mathbf{x} + \mathbf{U}^H \mathbf{w}$$

$$\tilde{\mathbf{y}} = \Sigma \mathbf{V}^H \mathbf{x} + \tilde{\mathbf{w}}, \text{ where } \tilde{\mathbf{y}} \triangleq \mathbf{U}^H \mathbf{y} \text{ and } \tilde{\mathbf{w}} \triangleq \mathbf{U}^H \mathbf{w}$$

Examples:

- I) SVD of $\mathbf{H} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$ is $= \begin{bmatrix} \frac{1}{\sqrt{2}} \\ \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} \sqrt{2} \end{bmatrix} \begin{bmatrix} 1 \end{bmatrix}$

where $\mathbf{U}^H \mathbf{U} = \mathbf{I}$, $\sigma_1 = \sqrt{2} > 0$ and $\mathbf{V} = [1]$ is unitary matrix

- II) $\mathbf{H} = \begin{bmatrix} 1 & 0 \\ 0 & \sqrt{5} \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} 0 & \sqrt{5} \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} \sqrt{5} & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$

where $\mathbf{U}^H \mathbf{U} = \mathbf{I}$, $\mathbf{V}^H \mathbf{V} = \mathbf{V} \mathbf{V}^H = \mathbf{I}$
 and $\sigma_1 = \sqrt{5} > \sigma_2 = \sqrt{1} > 0$

- III) $\mathbf{H} = \begin{bmatrix} 1 & 2 \\ 1 & -2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 0 & 2 \\ 1 & 0 \end{bmatrix}$
- $$= \begin{bmatrix} 1 & 1 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} \frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \\ -\frac{1}{\sqrt{2}} & \frac{1}{\sqrt{2}} \end{bmatrix} \begin{bmatrix} 2\sqrt{2} & 0 \\ 0 & \sqrt{2} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

Solution using Lagrange multipliers

- i) Need to maximize the below equation

$$\sum_{i=0}^t \log_2 \left(1 + \frac{\sigma_i^2 P_i}{\sigma^2} \right) + \lambda \left(P - \sum_{i=0}^t P_i \right)$$

- ii) From partial differentiation w.r.t P_i and equating to 0

$$\frac{\partial}{\partial P_i} = \frac{\sigma_i^2 / \sigma^2}{1 + \sigma_i^2 P_i / \sigma^2} + \lambda(0 - 1) = 0$$

$$\Rightarrow \lambda = \frac{\sigma_i^2 / \sigma^2}{1 + \sigma_i^2 P_i / \sigma^2}, \Rightarrow P_i = \left(\frac{1}{\lambda} - \frac{\sigma_i^2}{\sigma^2} \right)^+$$

$$P_i = \begin{cases} \frac{1}{\lambda} - \frac{\sigma_i^2}{\sigma^2} & \text{if } \frac{1}{\lambda} - \frac{\sigma_i^2}{\sigma^2} \geq 0, \\ 0 & \text{otherwise} \end{cases}$$

- iii) To obtain the value of Lagrange multiplier λ , need to solve

$$\sum_{i=0}^t P_i = P, \Rightarrow \sum_{i=0}^t \left(\frac{1}{\lambda} - \frac{\sigma_i^2}{\sigma^2} \right) = P$$

→ This algorithm is termed as the Water Filling Algorithm ■

Spatial Multiplexing

- i) At the tx, by performing precoding i.e., $\mathbf{x} = \mathbf{V} \tilde{\mathbf{x}}$, we have

$$\begin{bmatrix} \tilde{y}_1 \\ \tilde{y}_2 \\ \vdots \\ \tilde{y}_t \end{bmatrix} = \begin{bmatrix} \sigma_1 & 0 & \cdots & 0 \\ 0 & \sigma_2 & & \\ \vdots & & \ddots & 0 \\ 0 & 0 & \cdots & \sigma_t \end{bmatrix} \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \\ \vdots \\ \tilde{x}_t \end{bmatrix} + \begin{bmatrix} \tilde{w}_1 \\ \tilde{w}_2 \\ \vdots \\ \tilde{w}_t \end{bmatrix},$$

$$\Rightarrow \tilde{y}_i = \sigma_i \tilde{x}_i + \tilde{w}_i, \text{ for } 1 \leq i \leq t$$

- ii) This indicates, there are t parallel channels and t information symbols can be transmitted in parallel
 Using SVD, SPATIAL MULTIPLEXING is obtained !

Example: Consider 3×3 MIMO Channel

- i) Given \mathbf{H} , and observing that its columns are orthogonal

$$\mathbf{H} = \begin{bmatrix} 2 & -6 & 0 \\ 3 & 4 & 0 \\ 0 & 0 & 2 \end{bmatrix} = \begin{bmatrix} \frac{2}{\sqrt{52}} & \frac{-6}{\sqrt{52}} & 0 \\ \frac{3}{\sqrt{13}} & \frac{4}{\sqrt{13}} & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{13} & 0 & 0 \\ 0 & \sqrt{52} & 0 \\ 0 & 0 & 2 \end{bmatrix}$$

To get the singular values in decreasing order

$$= \begin{bmatrix} \frac{-6}{\sqrt{52}} & \frac{2}{\sqrt{13}} & 0 \\ \frac{4}{\sqrt{52}} & \frac{3}{\sqrt{13}} & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 0 & \sqrt{52} & 0 \\ \sqrt{13} & 0 & 0 \\ 0 & 0 & 2 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{-6}{\sqrt{52}} & \frac{2}{\sqrt{13}} & 0 \\ \frac{4}{\sqrt{52}} & \frac{3}{\sqrt{13}} & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \sqrt{52} & 0 & 0 \\ 0 & \sqrt{13} & 0 \\ 0 & 0 & 2 \end{bmatrix} = \mathbf{U} \Sigma \mathbf{V}^H$$

- Note, $\mathbf{U}^H \mathbf{U} = \mathbf{I}$, $\mathbf{V}^H \mathbf{V} = \mathbf{V} \mathbf{V}^H = \mathbf{I}$ and $\sigma_1 = \sqrt{52} > \sigma_2 = \sqrt{13} > \sigma_3 = 2 > 0$

- ii) Applying transmit preprocessing (Precoding) at the transmitter,

$$\mathbf{x} = \mathbf{V} \tilde{\mathbf{x}}, \text{ i.e., } \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \\ \tilde{x}_3 \end{bmatrix}$$

- iii) At the receiver, $\tilde{\mathbf{y}} = \Sigma \tilde{\mathbf{x}} + \tilde{\mathbf{w}}$

$$\text{i.e., } \begin{bmatrix} \tilde{y}_1 \\ \tilde{y}_2 \\ \tilde{y}_3 \end{bmatrix} = \begin{bmatrix} \sqrt{52} & 0 & 0 \\ 0 & \sqrt{13} & 0 \\ 0 & 0 & 2 \end{bmatrix} \begin{bmatrix} \tilde{x}_1 \\ \tilde{x}_2 \\ \tilde{x}_3 \end{bmatrix} + \begin{bmatrix} \tilde{w}_1 \\ \tilde{w}_2 \\ \tilde{w}_3 \end{bmatrix}$$

→ 3 decoupled channels resulting in spatial multiplexing.

Example: Optimal Power Allocation

- i) To achieve Shannon capacity

$$P_1 = \left(\frac{1}{\lambda} - \frac{\sigma_1^2}{\sigma^2} \right)^+ = \left(\frac{1}{\lambda} - \frac{\sigma_1^2}{52} \right)^+, P_2 = \left(\frac{1}{\lambda} - \frac{\sigma_2^2}{\sigma^2} \right)^+ = \left(\frac{1}{\lambda} - \frac{\sigma_2^2}{13} \right)^+,$$

$$\text{and } P_3 = \left(\frac{1}{\lambda} - \frac{\sigma_3^2}{\sigma^2} \right)^+ = \left(\frac{1}{\lambda} - \frac{\sigma_3^2}{4} \right)^+$$

- ii) Considering $\sigma^2 = 0dB = 1$ and total Tx power $P = 3dB \approx 2$

$$\rightarrow P_1 + P_2 + P_3 = 2, \Rightarrow \frac{1}{\lambda} = 0.7821$$

$$P_1 = -1.1755dB, P_2 = -1.517dB \text{ and } P_3 = -2.74dB$$

→ Power allocation decreases as the channel gain σ_i^2 decreases

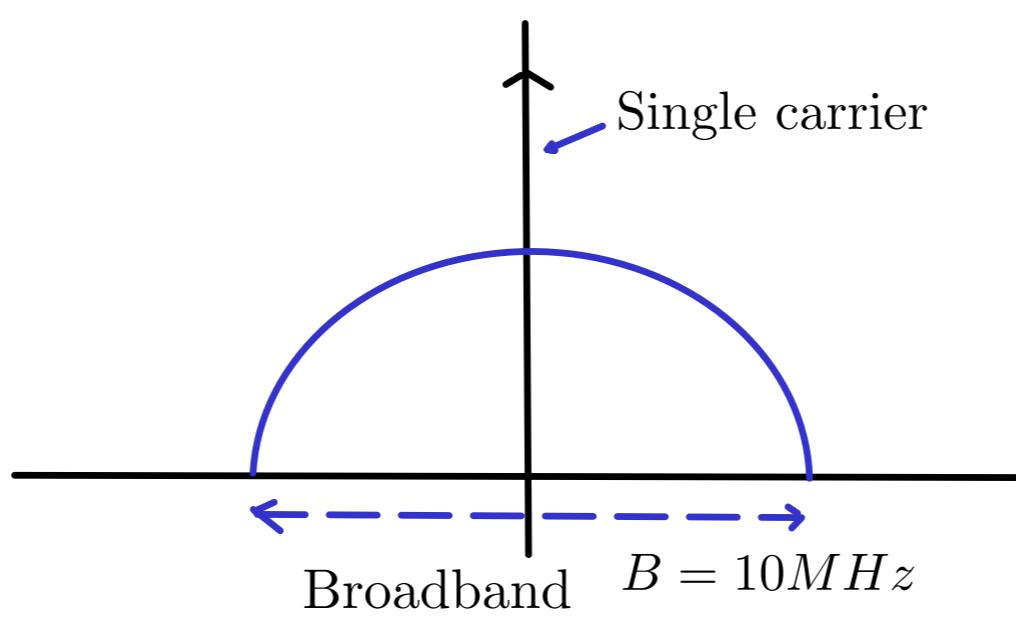
Orthogonal Frequency Division Multiplexing

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Orthogonal Frequency Division Multiplexing

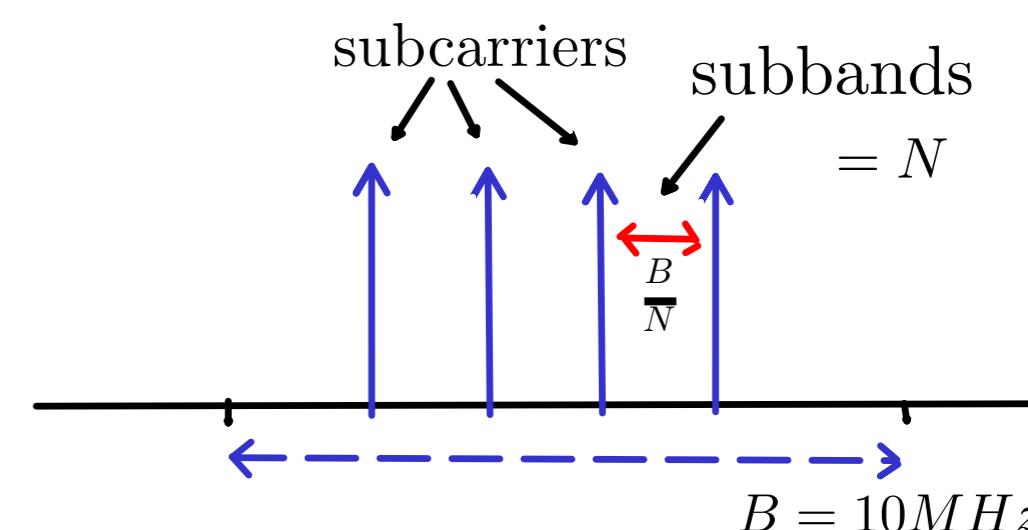
- i) Key enabler of high data rates (BROADBAND Technology) in 4G wireless systems and WiFi (WLAN) systems
 - LTE: Long Term Evolution
 - LTE-A: LTE Advanced
 - WiMAX: Worldwide Interoperability for Microwave Access
 - WLAN: 802.11a, 802.11g, 802.11n, 802.11ac

Basic Principle of OFDM



- i) Consider a broadband channel with single-carrier and bandwidth $B = 10 \text{ MHz}$,
- Symbol duration $T = \frac{1}{B} = \frac{1}{10 \text{ MHz}} = 0.1 \mu\text{s}$
- Typical delay spread of a channel $T_d \approx 2 - 3 \mu\text{s}$
⇒ $T \ll T_d$ (i.e., Symbol Time ≪ Delay spread), results in ISI or Inter symbol interference
- As bandwidth B increases symbol time decreases and leads to ISI
- ii) It is a significant challenge in a broadband wireless system to overcome ISI

Principle to Overcome ISI



- i) Divide the bandwidth B into N subbands of bandwidth $\frac{B}{N}$ each
Use a sub-carrier for each subband
- ii) Example: $B = 10 \text{ MHz}$ and $N = 1000$
⇒ BW of each subband = $\frac{10 \text{ MHz}}{1000} = 10 \text{ KHz}$
Symbol time in each subband $T = \frac{1}{10 \text{ KHz}} = 100 \mu\text{s} \gg T_d (= 2 \mu\text{s})$
⇒ NO ISI in new system with Multi-Carrier Modulated (MCM) system

Alleviate the need for large number of subcarriers

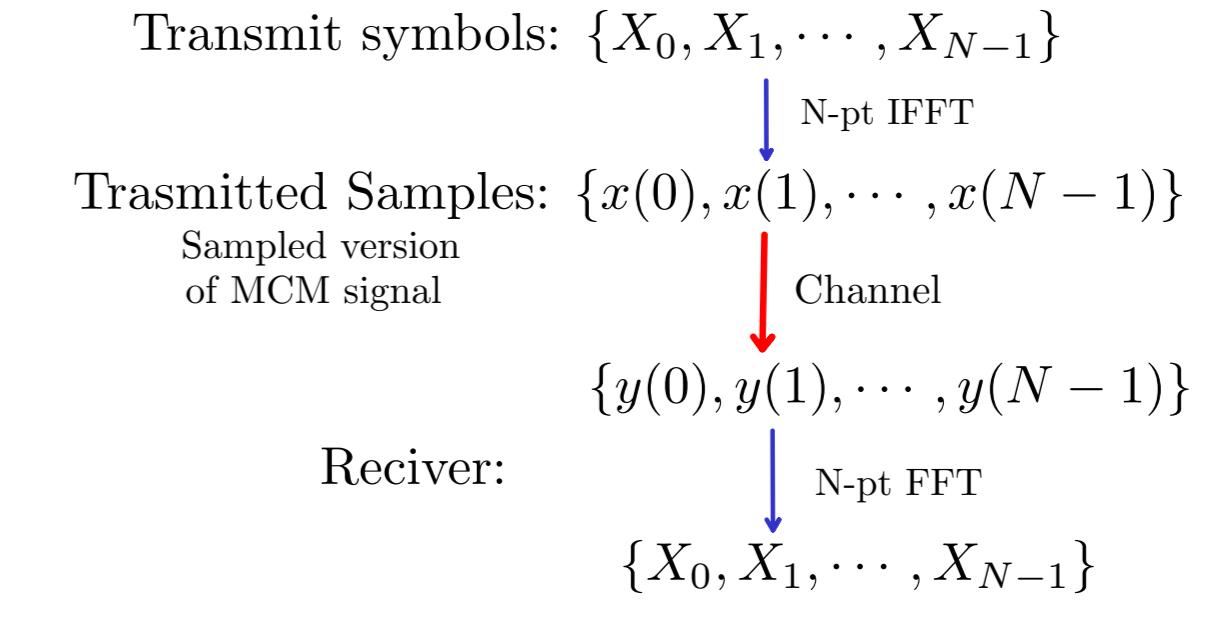
- i) Since the signal is bandlimited to $\{-f_{max}, f_{max}\}$ i.e., $\{-\frac{B}{2}, \frac{B}{2}\}$, it can be sampled at the Nyquist rate = $2f_{max} = B$
- ii) Sampling interval = $\frac{1}{B} = T$
 l^{th} sampling instant = $lT = \frac{l}{B}$
- iii) we have

$$x(t) = \sum_k X_k e^{j2\pi k f_0 t}$$

$$x(l) = x(lT) = \sum_k X_k e^{j2\pi k f_0 lT} = \sum_k X_k e^{j2\pi k \frac{B}{N} l \frac{l}{B}}$$

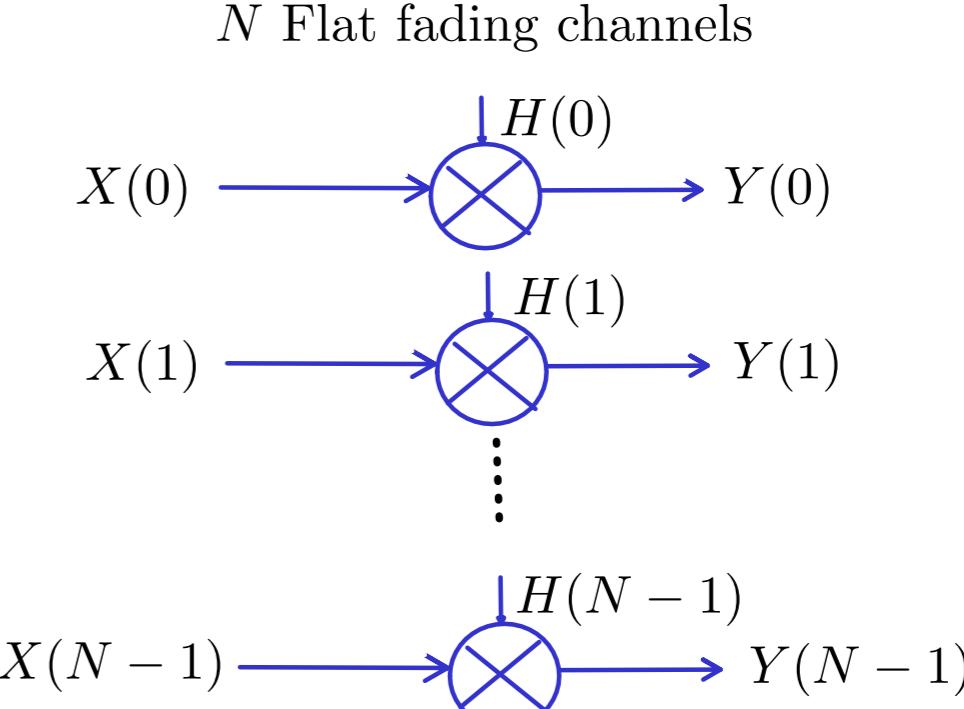
$$= \sum_k X_k e^{j2\pi k \frac{l}{N}}$$
 (which is IDFT or IFFT)
- where $x(l)$ is the l^{th} IDFT point of symbols transmitted on N subcarriers $\{X_0, X_1, \dots, X_{N-1}\}$
- iv) The principle of OFDM is to transmit the sample points i.e., IFFT at Transmitter and FFT at Receiver

IFFT at Transmitter and FFT at Receiver



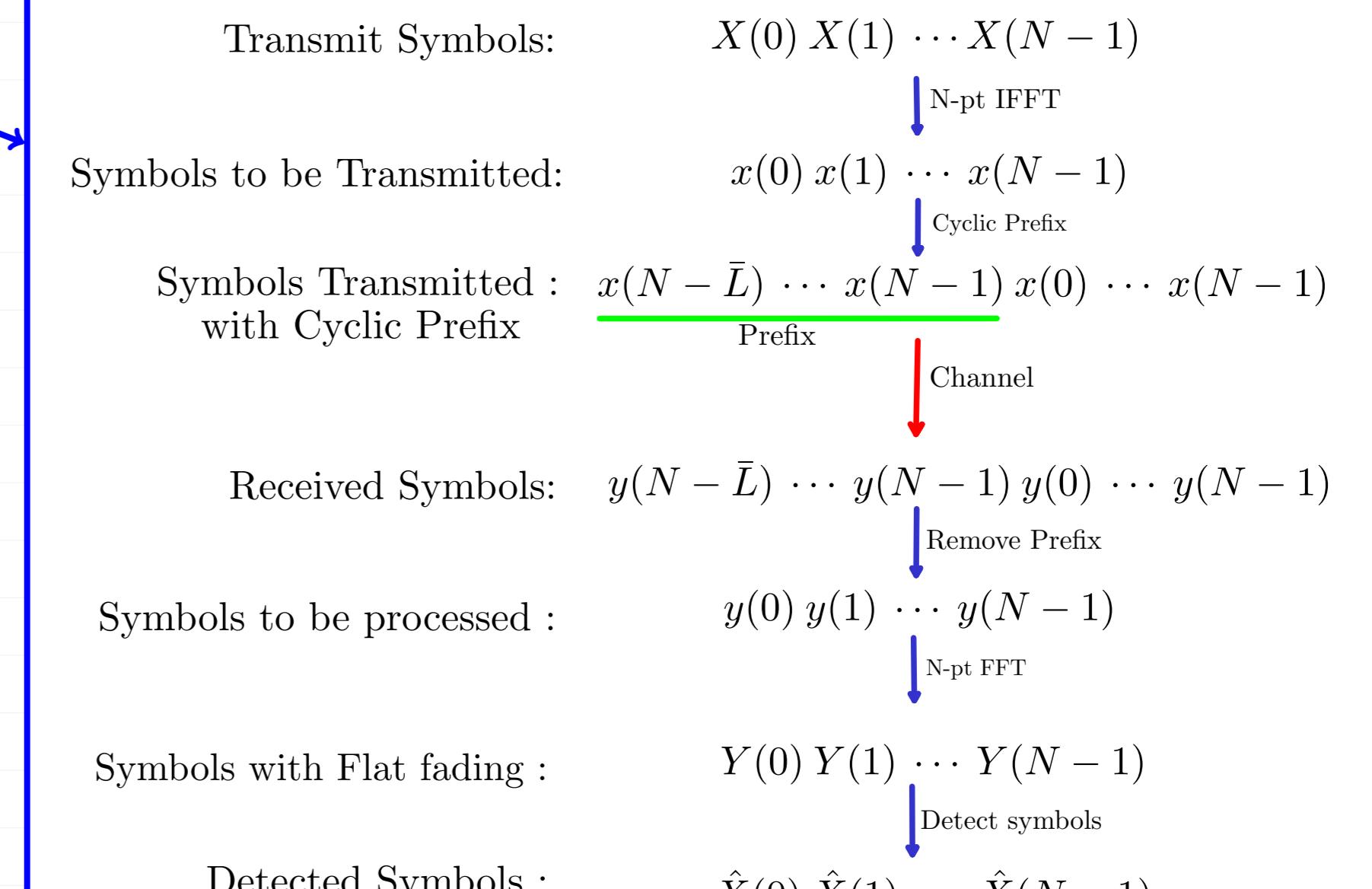
Equivalent N Flat fading channels

- i) Equivalently, $Y(k)$ can be considered received symbol on k^{th} subcarrier, $H(k)$ channel coefficient of k^{th} subcarrier and $X(k)$ as the transmitted symbol on the k^{th} subcarrier



⇒ OFDM has converted Frequency Selective channel into equivalent N Flat fading channels

Tx and Rx Schematic



Example:

- Q) Given $L = 16$ channel taps, $N = 256$ subcarriers and $\text{SNR} = 35 \text{ dB}$, compute OFDM BER.

$$\text{Sol) } \text{BER}_{\text{OFDM}} = \frac{1}{2} \left(1 - \sqrt{\frac{\frac{16}{256} 10^{3.5}}{2 + \frac{16}{256} 10^{3.5}}} \right) = 2.5 \times 10^{-3}$$

References

- [1] NOC: Principles of Modern CDMA MIMO OFDM Wireless Communications By Prof. Aditya K. Jagannatham, IIT Kanpur