

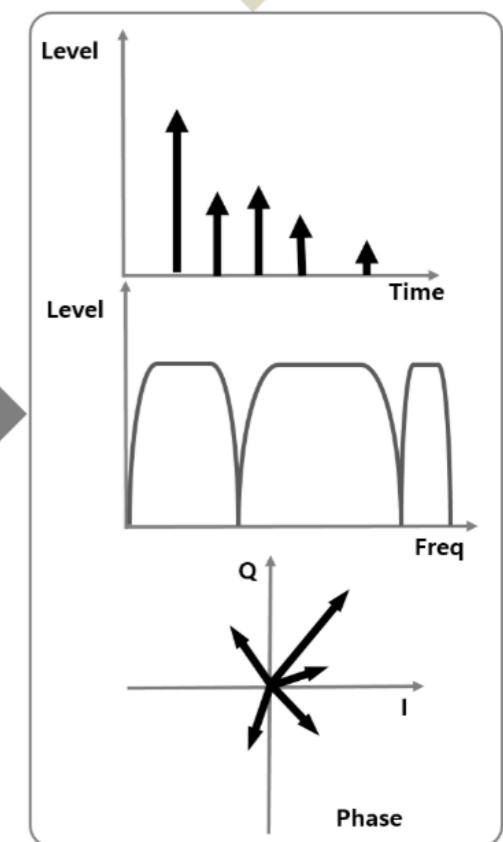
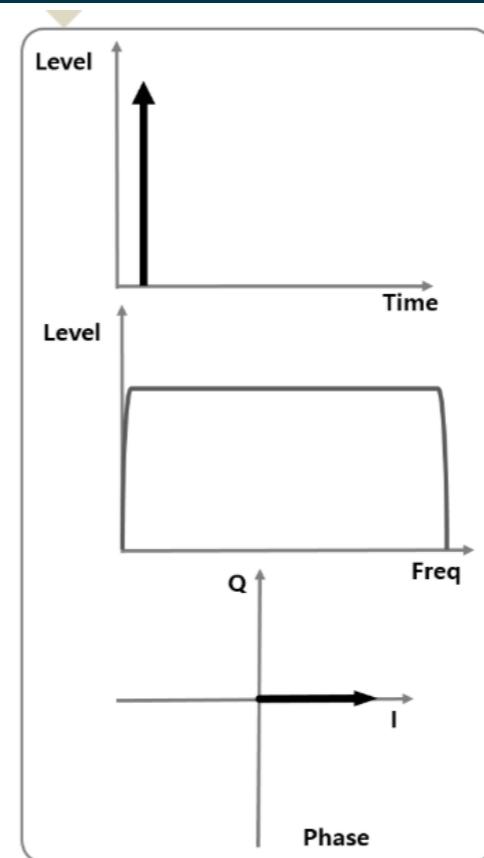
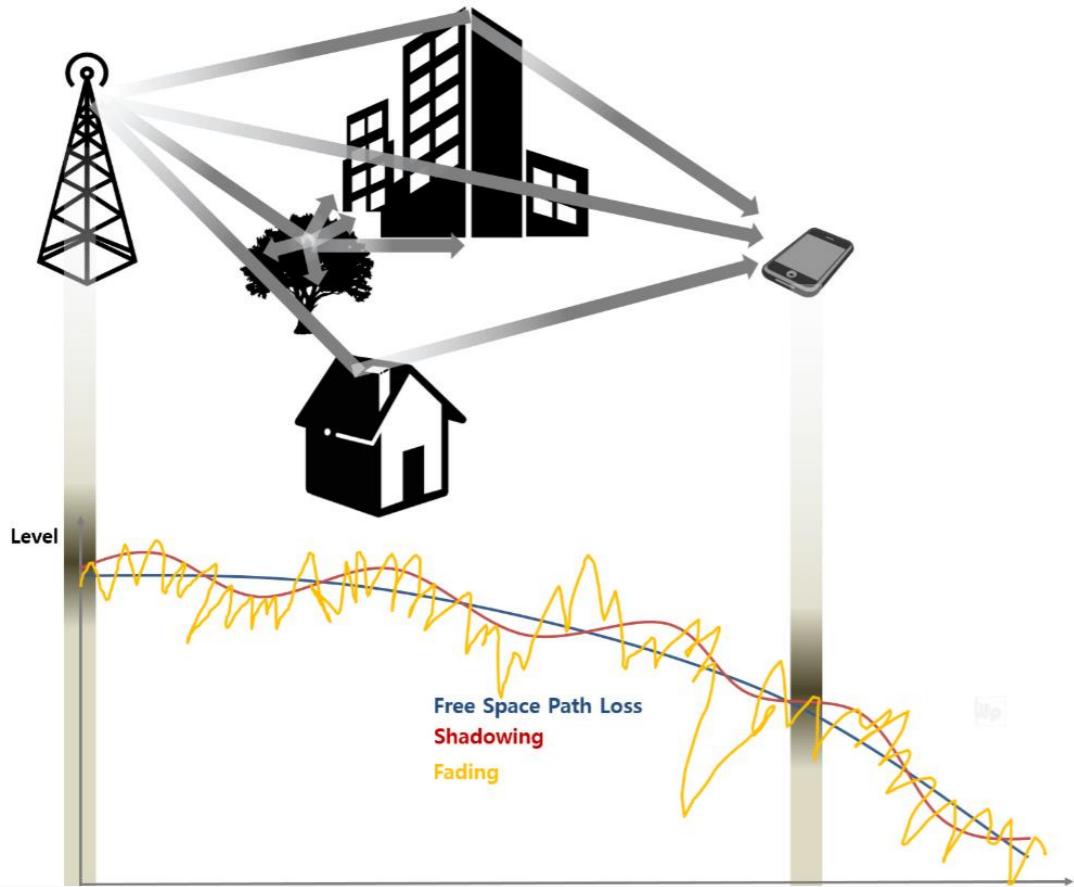
Lecture 2

Wireless Channel

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3/9, 2021

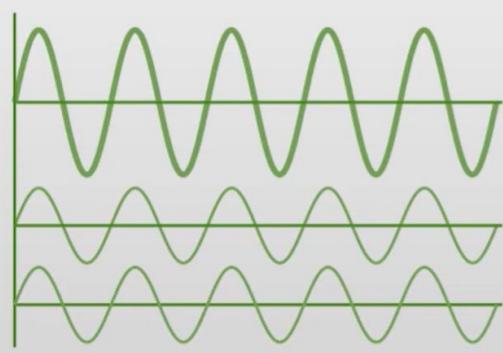
Major effect of the wireless channel



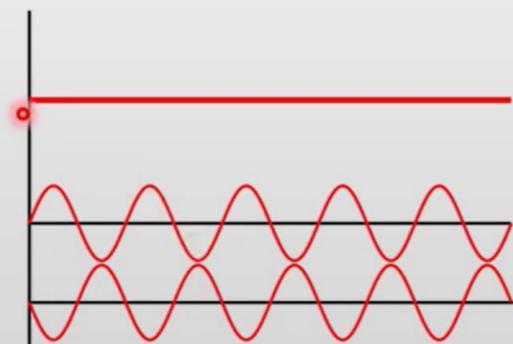
Signal in Wireless Channel

Effects of Multipath Propagation

Radio Waves reaching the receiver in more than one direction may
Constructive or **Destructive**



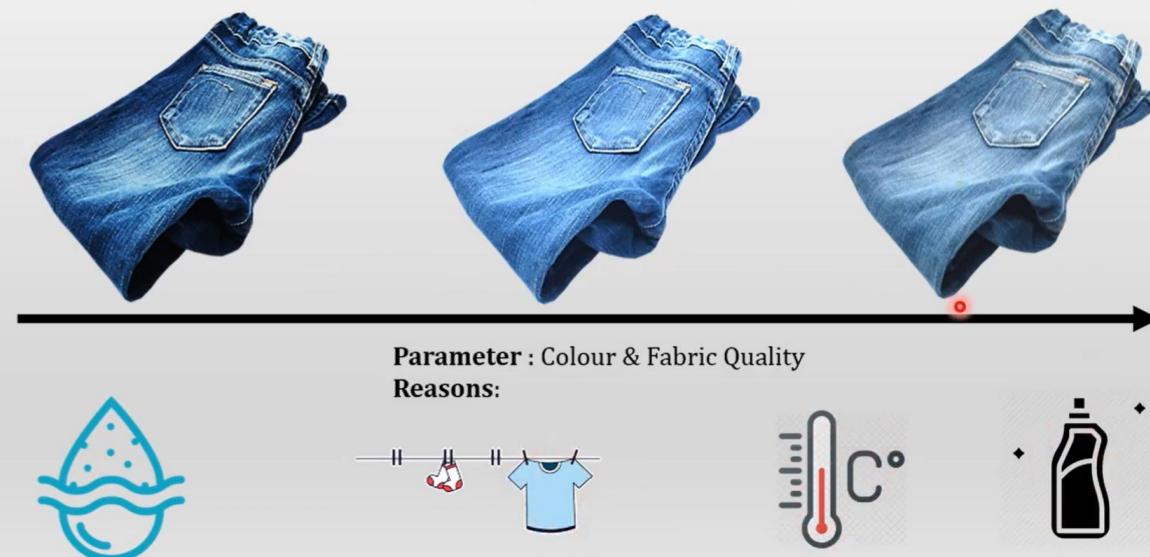
CONSTRUCTIVE SCENARIO



DESTRUCTIVE SCENARIO

Fading - Example

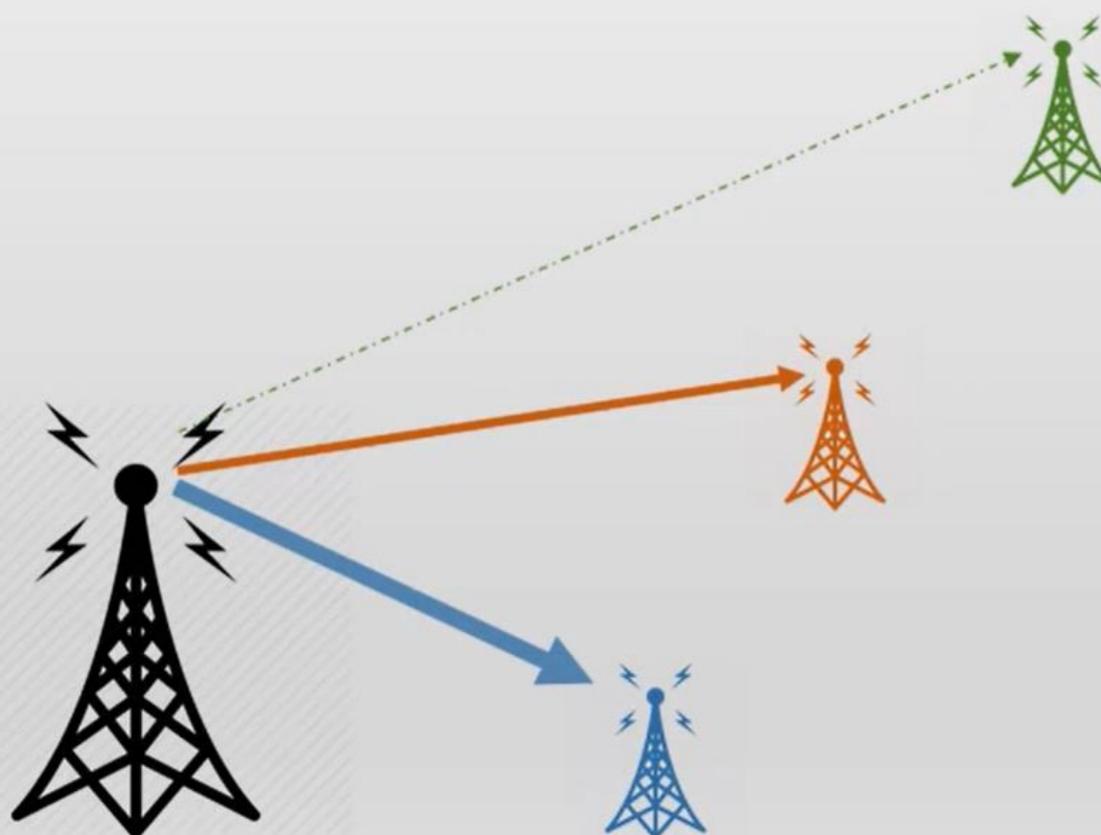
Fading in general is referred to as **Gradual Loss of Certain Quality / Parameter** and merely disappears over the period of time.



Fading

Fading

Fading is variation of the **attenuation of a signal** with various variables. These variables include **time**, **geographical position**, and **radio frequency**. In wireless systems, fading may either be due to multipath propagation, weather (particularly rain)



PATH LOSS



RAIN



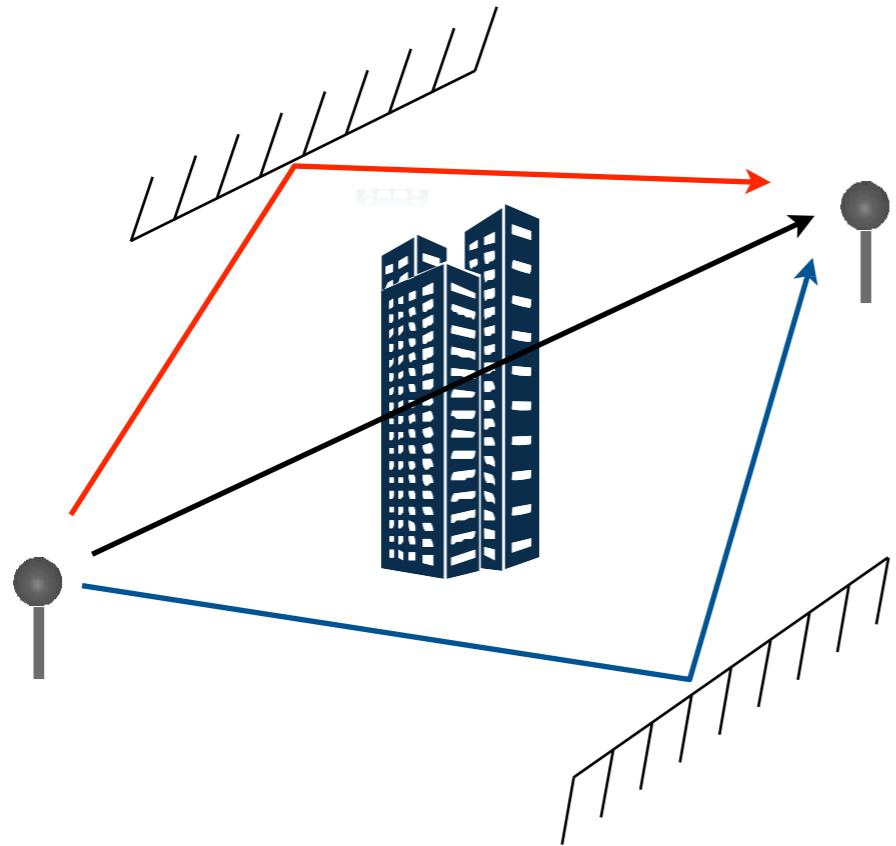
OBSTACLES



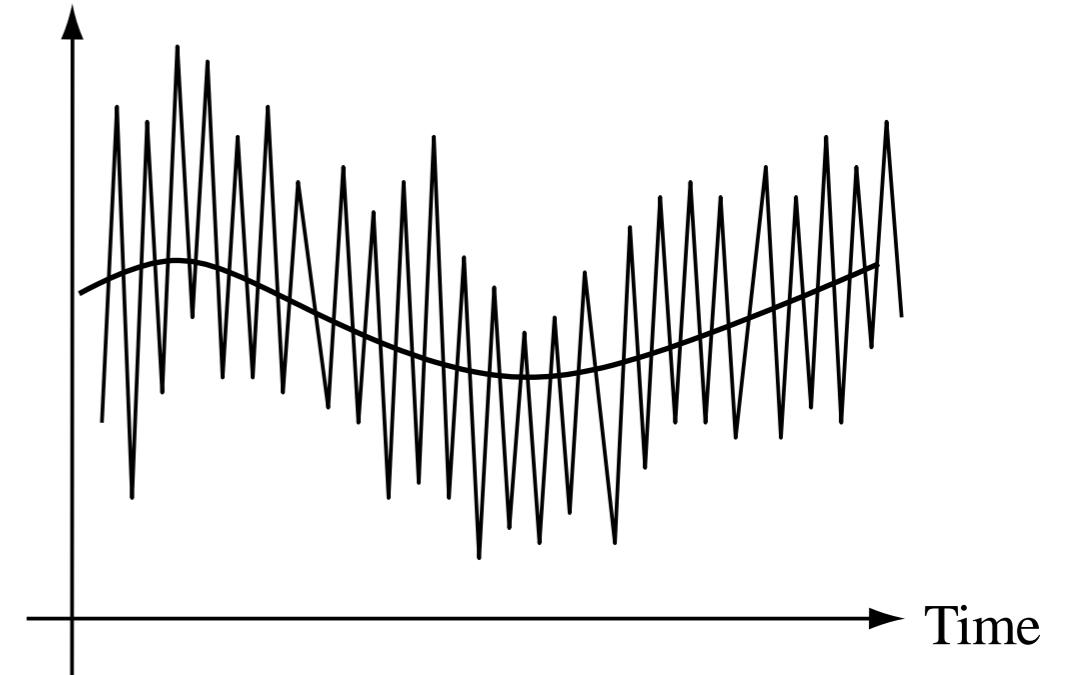
FREQUENCY



Wireless channels vary at two scales



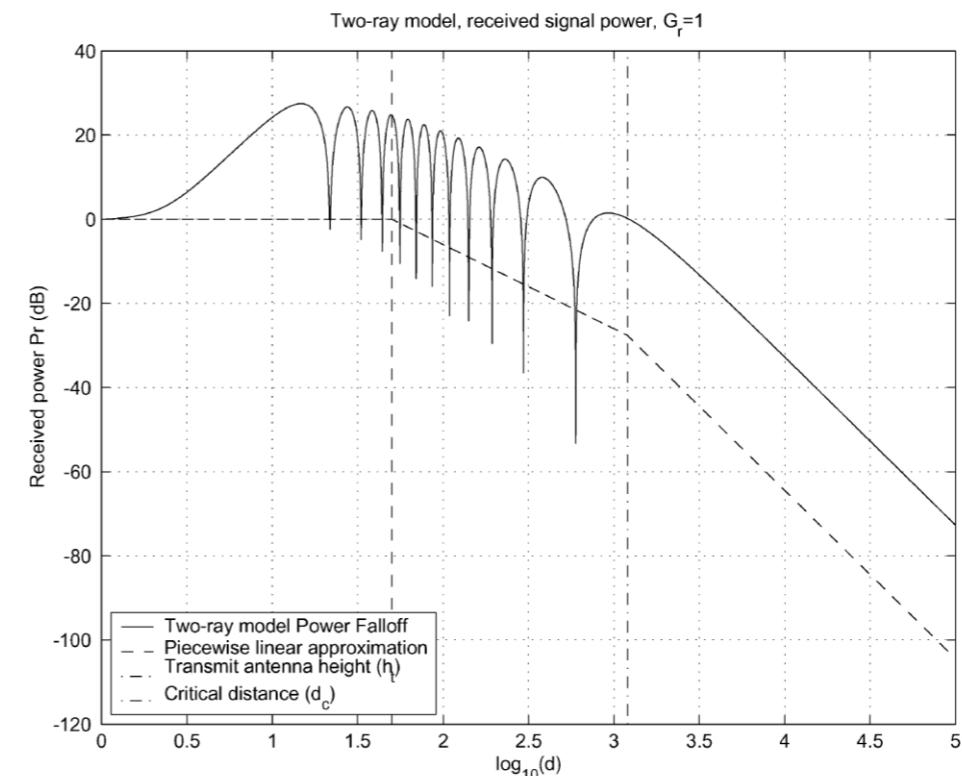
Channel quality



- Large-scale fading: path loss, shadowing, etc.
- Small-scale fading: constructive/destructive interference

Large - Scale Fading

- Path loss and Shadowing
 - In free space, received power $\propto \frac{1}{r^2}$
 - With reflections and obstacles, can attenuate faster than $\frac{1}{r^2}$
- Variation over time: very slow, order of seconds
- Critical for coverage and cell-site planning



Small - Scale Fading

- Multipath fading: due to constructive and destructive interference of the waves
- Channel varies when the mobile moves a distance of the order of the carrier wavelength λ
 - Typical carrier frequency $\sim 1\text{GHz} \Rightarrow \lambda = c/f = 0.3m$
- Variation over time: order of hundreds of microseconds
- Critical for design of communication systems

Plan

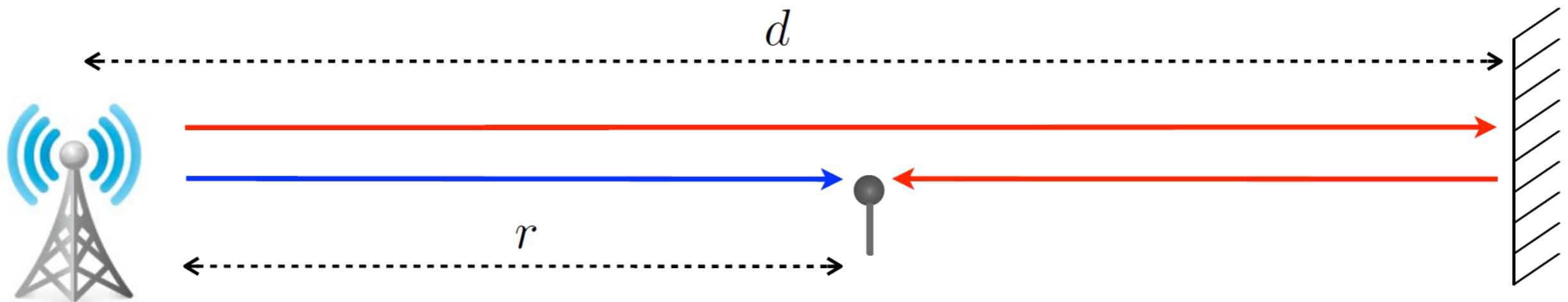
- Understand how *physical parameters* impact a wireless channel from the **communication system** point of view.
Physical parameters such as
 - Carrier frequency
 - Mobile speed
 - Bandwidth
 - Delay spread
 - etc.
- Start with **deterministic physical models**
- Progress towards **statistical** models

Outline

- Physical modeling of wireless channels
- Deterministic Input-output model
- Time and frequency coherence
- Statistical models

Physical Model: Warm-up Examples

Physical Model: Simple Example 1



Transmitted Waveform (electric field): $\cos 2\pi f t$

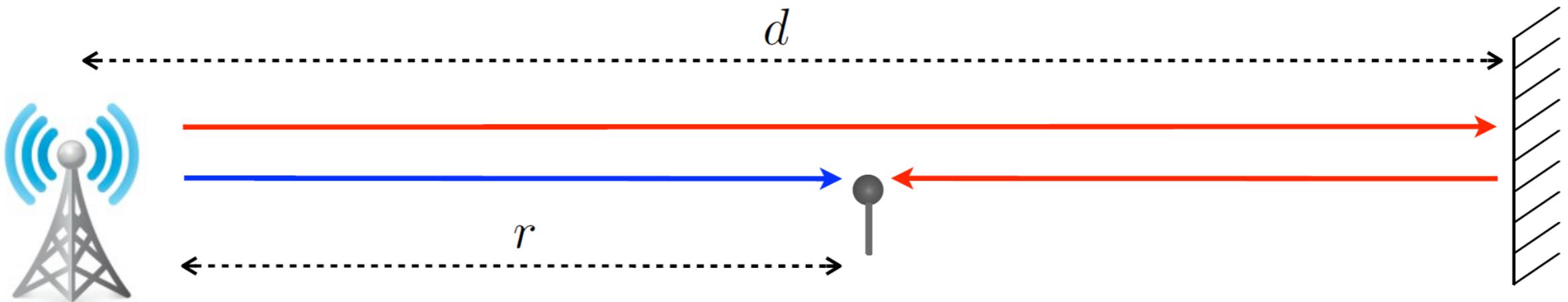
Received Waveform (path 1): $\frac{\alpha}{r} \cos 2\pi f \left(t - \frac{r}{c} \right)$

Received Waveform (path 2): $-\frac{\alpha}{2d - r} \cos 2\pi f \left(t - \frac{2d - r}{c} \right)$

\implies Received Waveform (aggregate):

$$\boxed{\frac{\alpha}{r} \cos 2\pi f \left(t - \frac{r}{c} \right) - \frac{\alpha}{2d - r} \cos 2\pi f \left(t - \frac{2d - r}{c} \right)}$$

Physical Model: Simple Example 1



Transmitted Waveform (electric field): $\cos 2\pi f t$

Received Waveform (aggregate):

$$\frac{\alpha}{r} \cos 2\pi f \left(t - \frac{r}{c} \right) - \frac{\alpha}{2d - r} \cos 2\pi f \left(t - \frac{2d - r}{c} \right)$$

Phase Difference between the two sinusoids:

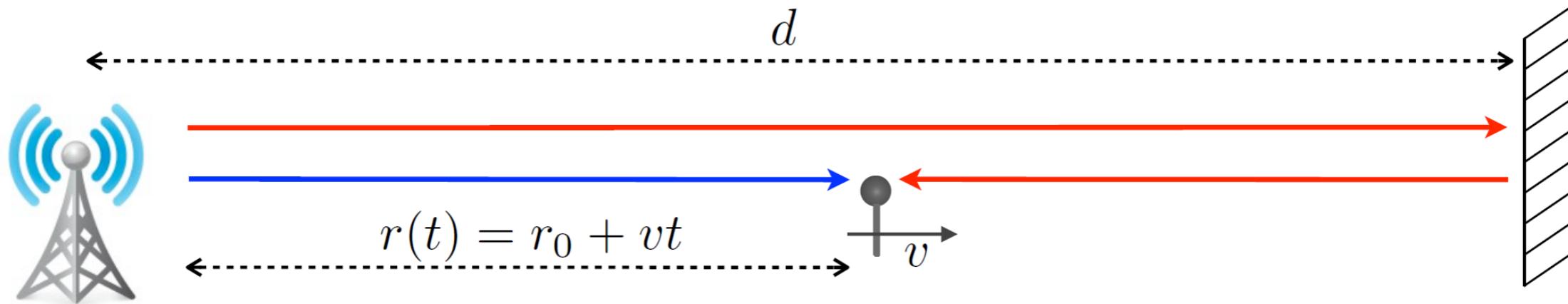
$$\begin{aligned}\Delta\theta &= \left\{ \frac{2\pi f(2d - r)}{c} + \pi \right\} - \frac{2\pi fr}{c} = 2\pi \frac{(2d - r) - r}{c} f + \pi \\ &= \begin{cases} 2n\pi, & \text{constructive interference} \\ (2n + 1)\pi, & \text{destructive interference} \end{cases}\end{aligned}$$

T_d
Delay Spread:
difference
between delays

Delay Spread and Coherence Bandwidth

- Delay spread T_d : difference between delays of paths
- If frequency f change by $1/(2T_d)$, then the combined received sinusoid move from peak to valley
- Therefore, the frequency-variation scale is of the order of $\frac{1}{T_d}$
- Coherence bandwidth $W_c := \frac{1}{T_d}$

Physical Model: Simple Example 2



Transmitted Waveform (electric field): $\cos 2\pi f t$

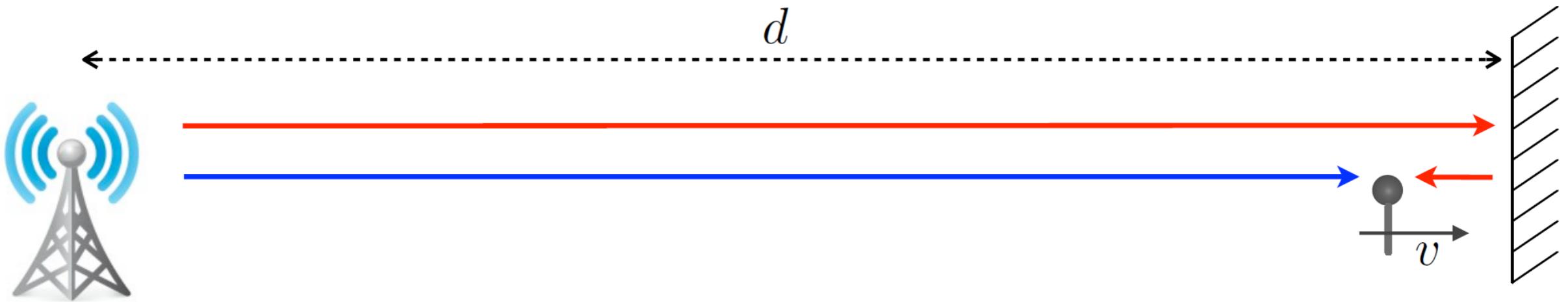
$$\text{Received Waveform (path 1): } \frac{\alpha}{r(t)} \cos 2\pi f \left(t - \frac{r(t)}{c} \right)$$

$$\text{Received Waveform (path 2): } -\frac{\alpha}{2d - r(t)} \cos 2\pi f \left(t - \frac{2d - r(t)}{c} \right)$$

⇒ Received Waveform (aggregate):

$$\begin{aligned} & \frac{\alpha}{r(t)} \cos 2\pi f \left(t - \frac{r(t)}{c} \right) - \frac{\alpha}{2d - r(t)} \cos 2\pi f \left(t - \frac{2d - r(t)}{c} \right) \\ &= \boxed{\frac{\alpha}{r_0 + vt} \cos 2\pi f \left[\left(1 - \frac{v}{c} \right) t - \frac{r_0}{c} \right] - \frac{\alpha}{2d - r_0 - vt} \cos 2\pi f \left[\left(1 + \frac{v}{c} \right) t - \frac{2d - r_0}{c} \right]} \end{aligned}$$

Physical Model: Simple Example 2



Approximation: distance to mobile Rx \ll distance to Tx

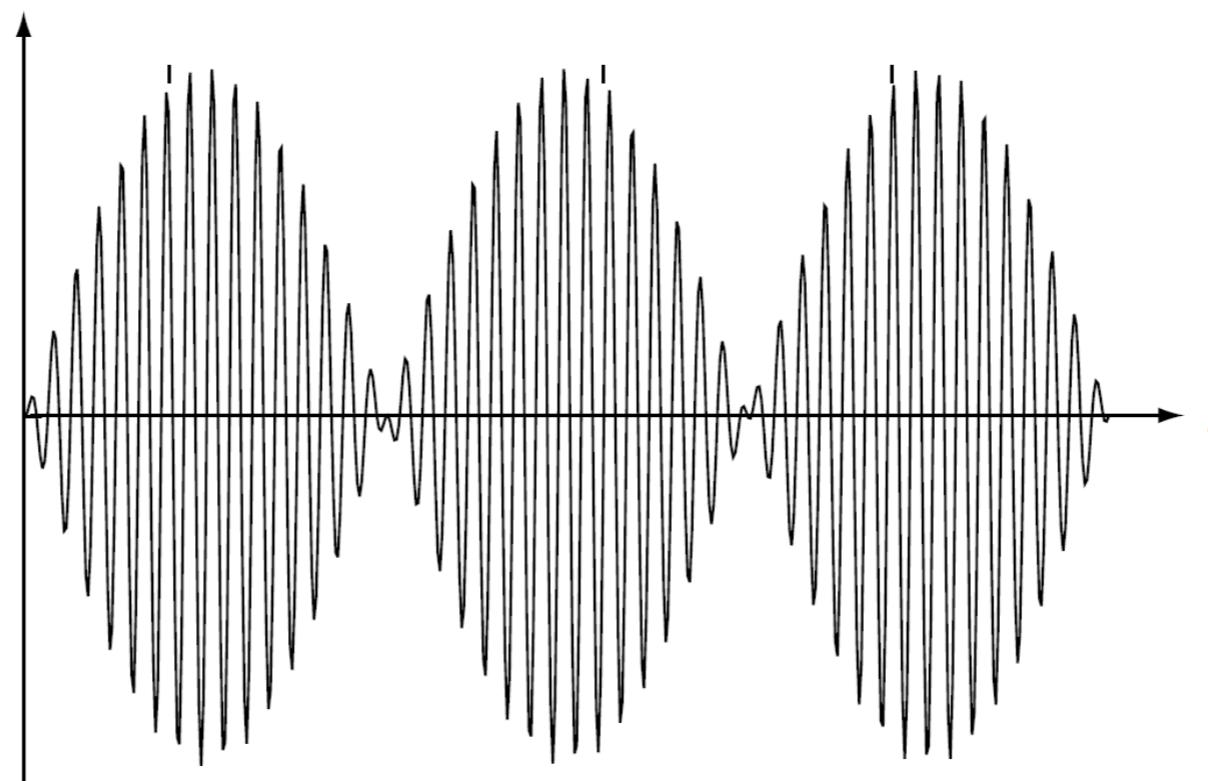
\implies Received Waveform (aggregate):

$$= \frac{\alpha}{r_0 + vt} \cos 2\pi f \left[\left(1 - \frac{v}{c}\right) t - \frac{r_0}{c} \right] - \frac{\alpha}{2d - r_0 - vt} \cos 2\pi f \left[\left(1 + \frac{v}{c}\right) t - \frac{2d - r_0}{c} \right]$$
$$\approx \boxed{\frac{2\alpha}{r_0 + vt} \sin 2\pi f \left(\frac{vt}{c} + \frac{r_0 - d}{c} \right)} \boxed{\sin 2\pi f \left(t - \frac{d}{c} \right)}$$

Time-varying amplitude

Time-invariant shift
of the original input
waveform

Physical Model: Simple Example 2



Time-varying envelope

$$\frac{2\alpha}{r_0 + vt} \sin 2\pi f \left(\frac{vt}{c} + \frac{r_0 - d}{c} \right)$$

Time-variation scale: r_0/v
(seconds or minutes),
much smaller than that of
the second term

Doppler Spread $D_s = \frac{2fv}{c}$

Difference of the Doppler shifts of
the two paths, cause this variation
over time.

Time-variation scale: c/fv (ms)

Doppler Spread and Coherence Time

- Mobility causes time-varying delays (Doppler shift)
- Doppler spread D_s : difference between Doppler shifts of multiple signal paths
- If time t change by $1/(2D_s)$, then the combined received sinusoidal envelope move from peak to valley
- Therefore, the time-variation scale is of the order of $\frac{1}{D_s}$
- Coherence time $T_c := \frac{1}{D_s}$

What we learned from the examples

- Delay spread/coherence bandwidth and Doppler spread/coherence time seem fundamental
- However, it is difficult to derive the explicit received waveform mathematically.
 - Out of scope – EM wave theory
- Instead, we construct useful input/output models, and take measurements to determine the parameters in the models

Physical Model: Input/Output Relations

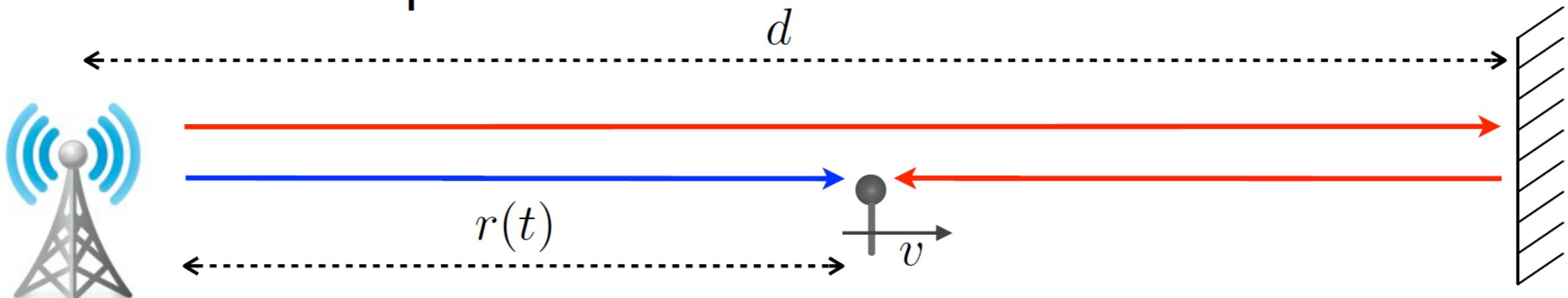
Physical Input/Output Model

- Wireless channels as linear time-varying systems:

$$y(t) = \sum_i a_i(t)x(t - \tau_i(t))$$

$a_i(t)$: gain of path i $\tau_i(t)$: delay of path i

- Recall Example 2:



$$x(t) = \cos 2\pi f t$$

$$a_1(t) = \frac{|\alpha|}{r_0 + vt}$$

$$a_2(t) = \frac{|\alpha|}{2d - r_0 - vt}$$

$$\tau_1(t) = \frac{r_0 + vt}{c}$$

$$\tau_2(t) = \frac{2d - r_0 - vt}{c} - \frac{\pi}{2\pi f}$$

Physical Input/Output Model

- Wireless channels as linear time-varying systems:

$$y(t) = \sum_i a_i(t)x(t - \tau_i(t))$$

$a_i(t)$: gain of path i $\tau_i(t)$: delay of path i

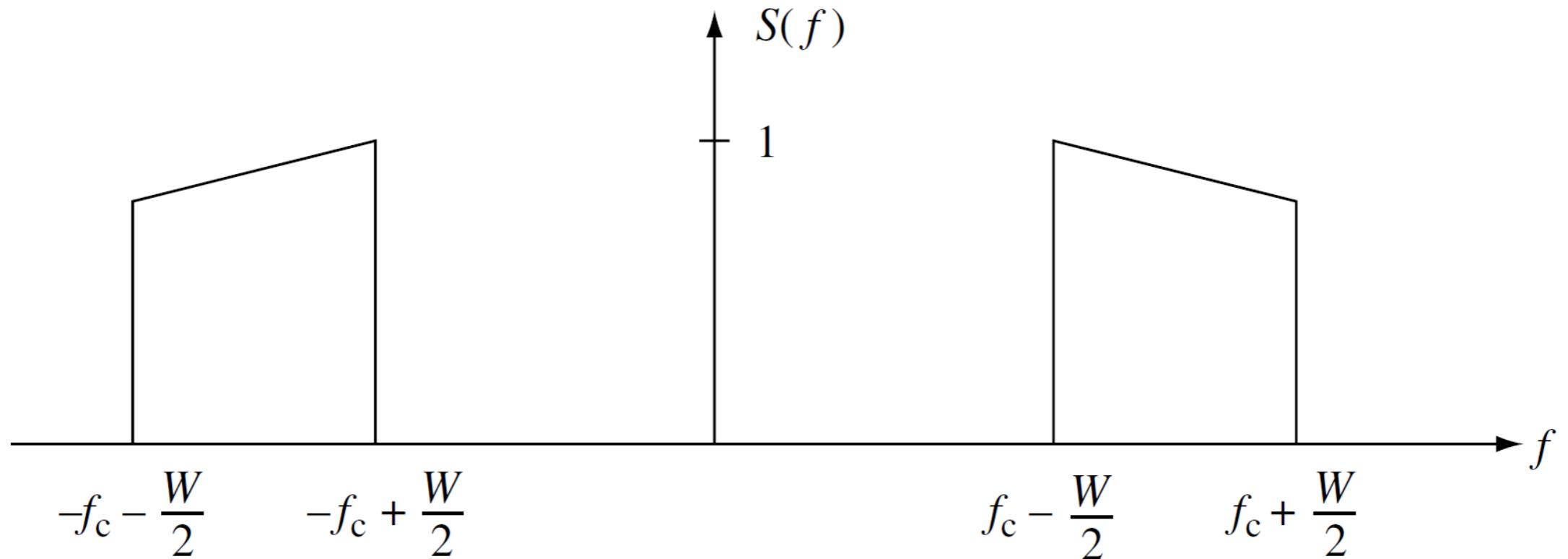
- Impulse response:

$$x(t) \rightarrow \boxed{h(\tau, t)} \rightarrow y(t) = \sum_i a_i(t)x(t - \tau_i(t))$$

$$h(\tau, t) = \sum_i a_i(t)\delta(\tau - \tau_i(t))$$

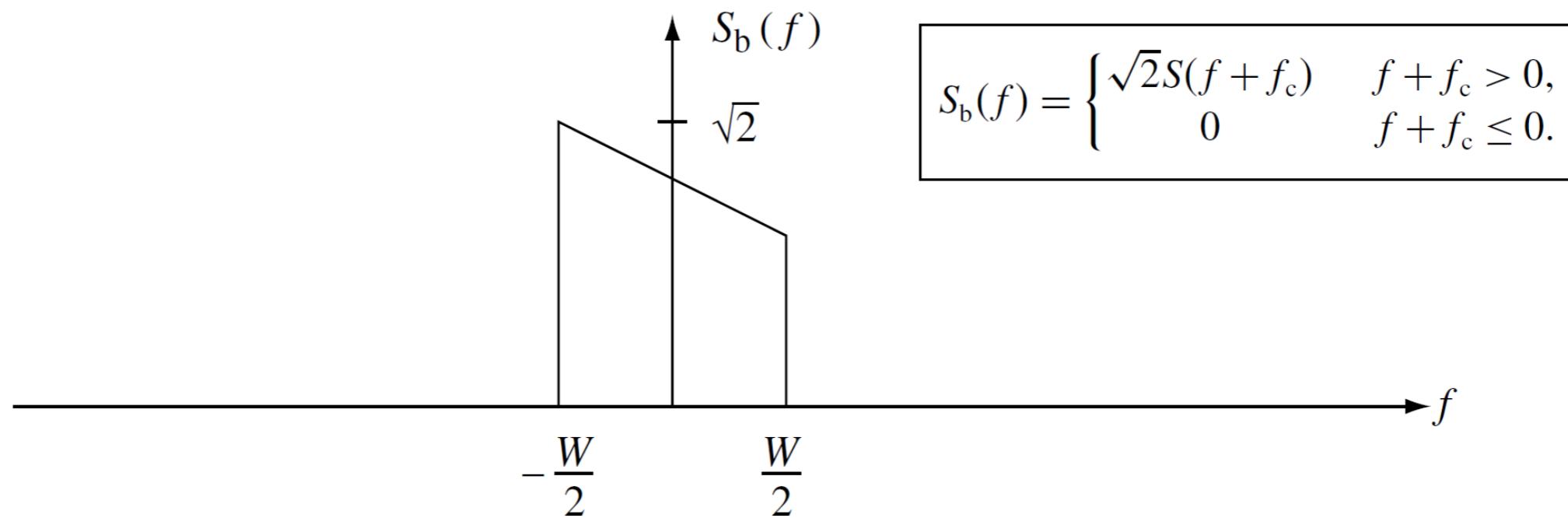
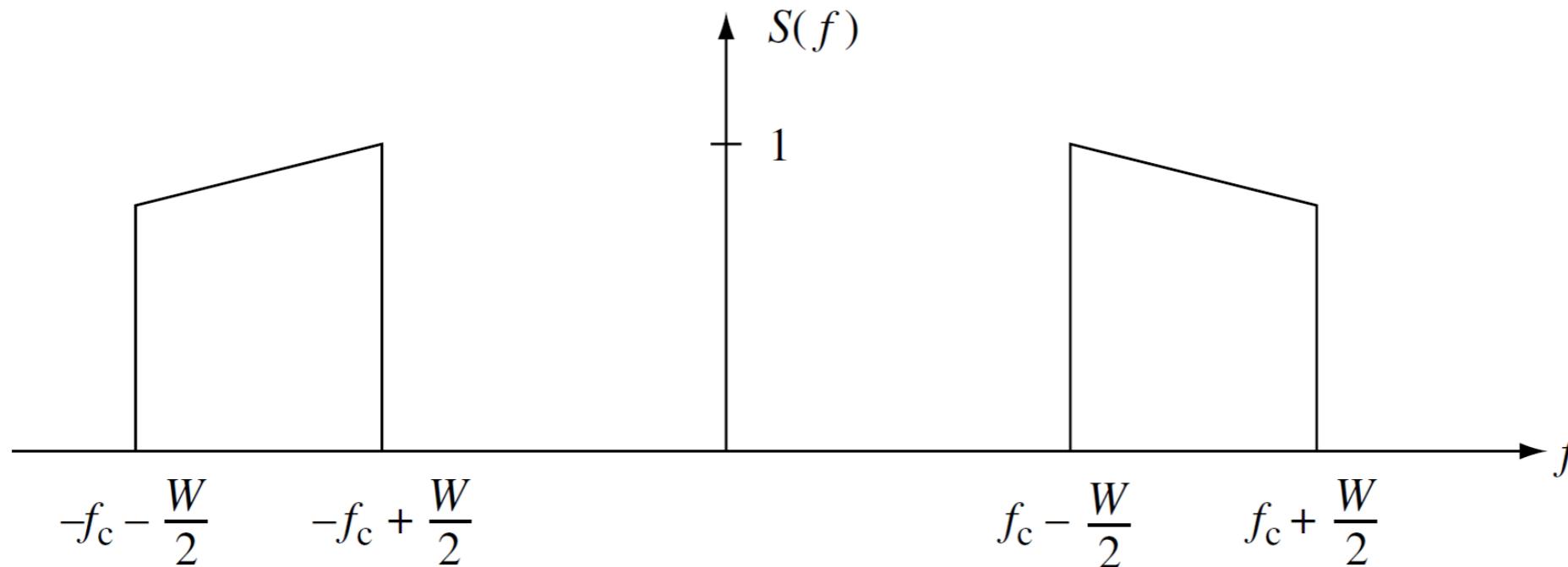
- Frequency response: $H(f; t) = \sum_i a_i(t)e^{-j2\pi f \tau_i(t)}$

Passband–Baseband Conversion

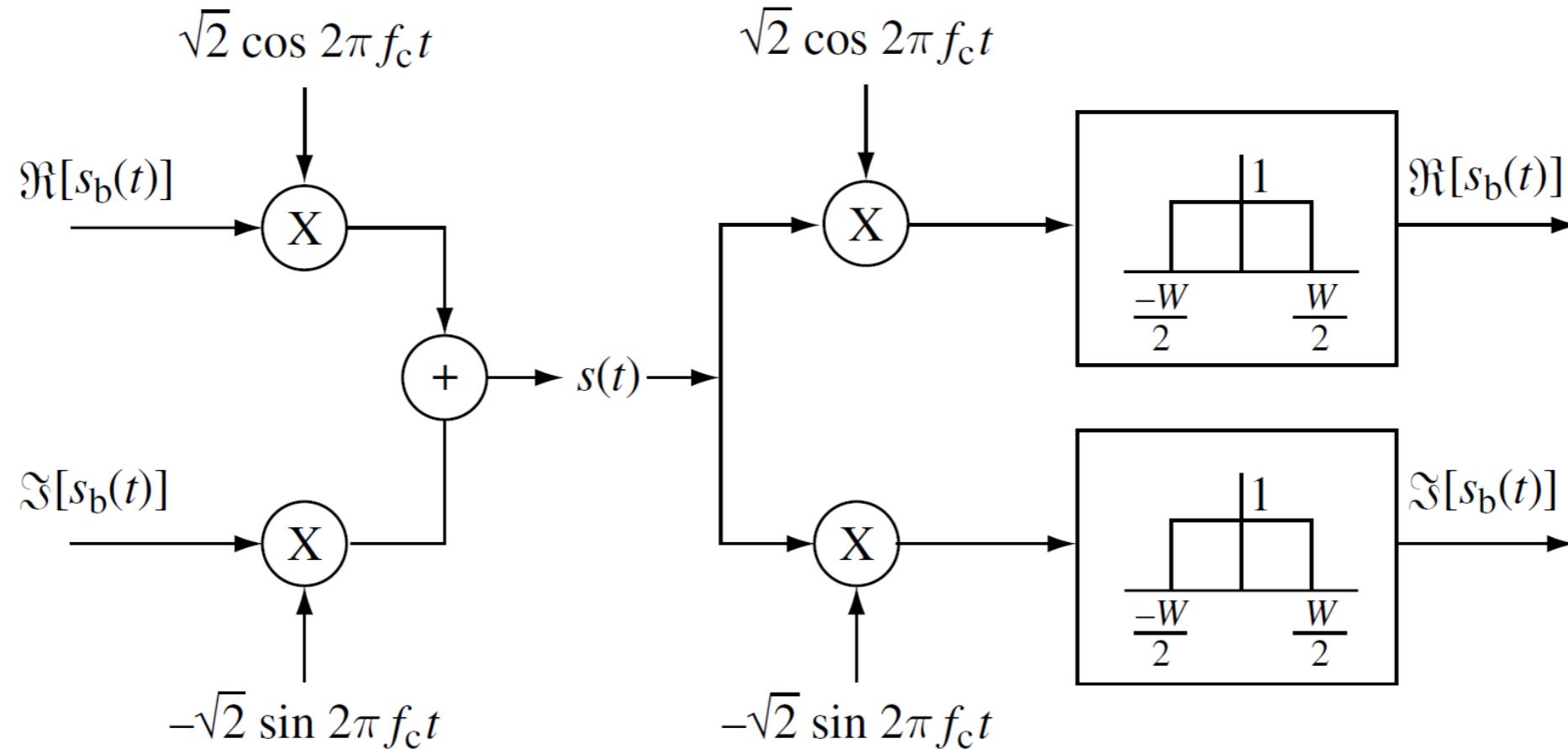


- Communications takes place in a passband
 - Carrier frequency f_c
 - Bandwidth $W < 2f_c$
 - Real signal $s(t)$

Passband–Baseband Conversion

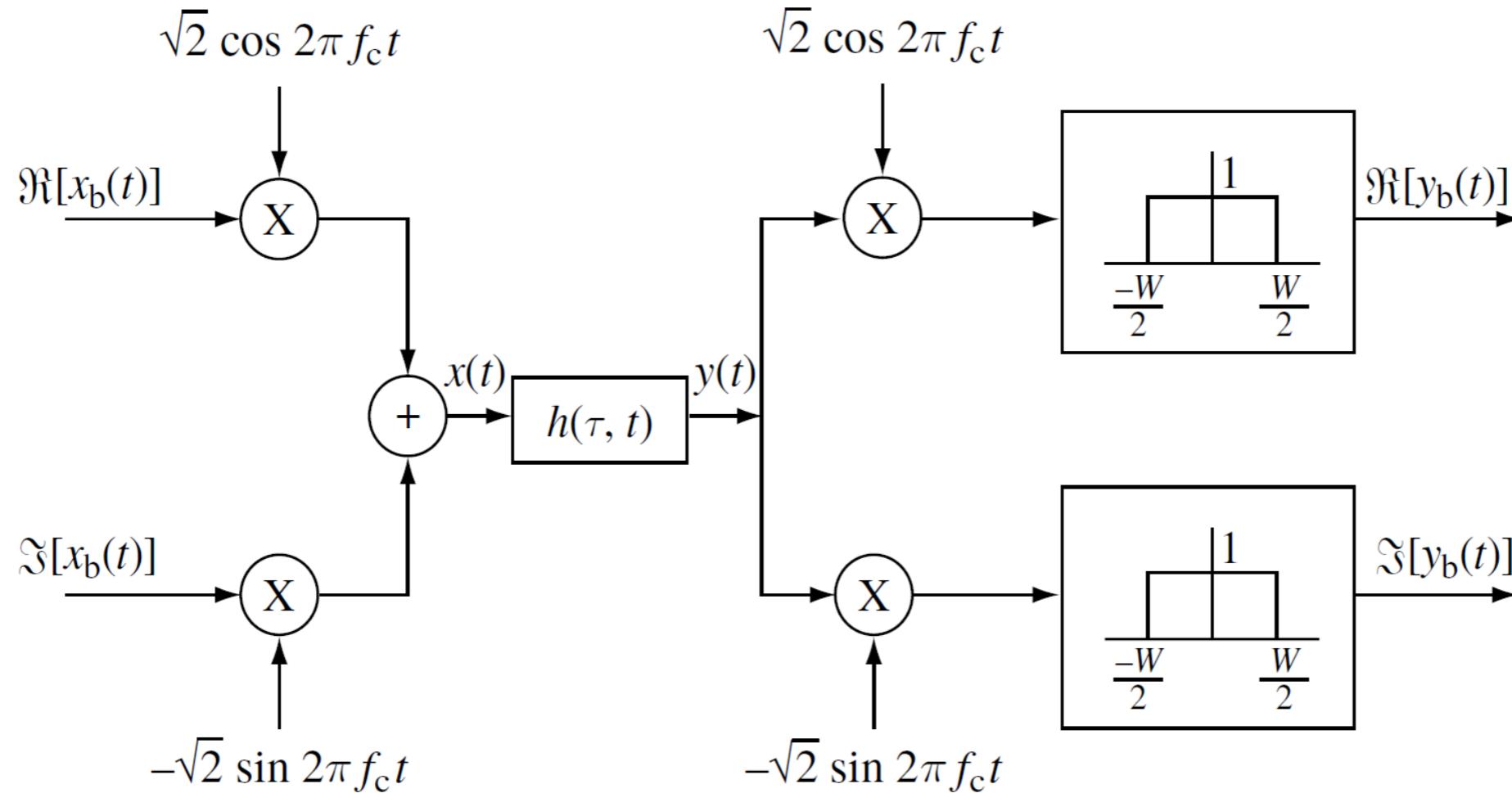


Passband–Baseband Conversion



$$s(t) = \frac{1}{\sqrt{2}} [s_b(t)e^{j2\pi f_c t} + s_b^*(t)e^{-j2\pi f_c t}] = \sqrt{2}\Re[s_b(t)e^{j2\pi f_c t}]$$

Baseband System Architecture



$$y_b(t) = \sum_i a_i^b(t) x_b(t - \tau_i(t)),$$

where $a_i^b(t) := a_i(t) e^{-j2\pi f_c \tau_i(t)}$

Continuous - time Baseband Model

- Complex baseband equivalent channel:

$$x_b(t) \rightarrow \boxed{h_b(\tau, t)} \rightarrow y_b(t) = \sum_i a_i^b(t) x_b(t - \tau_i(t))$$

$$h_b(\tau, t) = \sum_i a_i^b(t) \delta(\tau - \tau_i(t)),$$

$$\text{where } a_i^b(t) := a_i(t) e^{-j2\pi f_c \tau_i(t)}$$

- Frequency response: shifted from passband to baseband

$$H_b(f; t) = H(f + f_c; t)$$

- Each path is associated with a **delay** and a **complex gain**

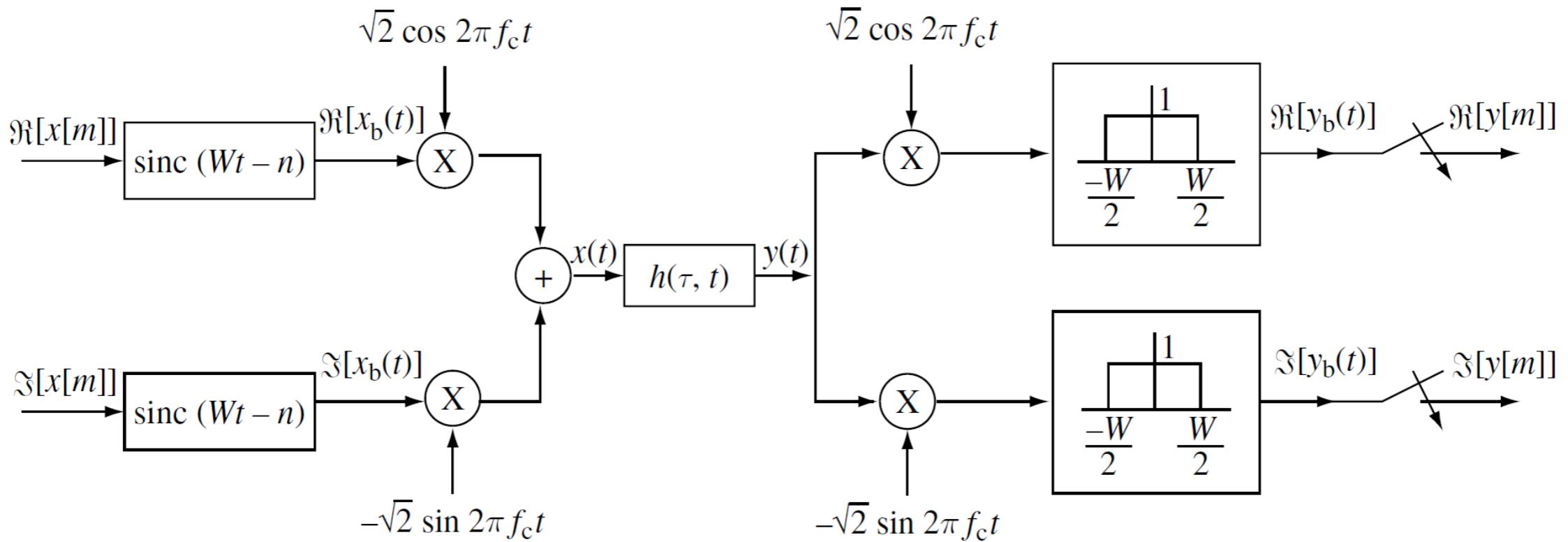
Modulation and Sampling

- Modern communication systems are digitized, (partially) thanks to **sampling theorem**
- Our baseband signal can be represented as follows:

$$x_b(t) = \sum_n x[n] \text{sinc}(Wt - n),$$

$$x[n] := x_n(n/W), \quad \text{sinc}(t) := \frac{\sin \pi t}{\pi t}$$

Modulation and Sampling



$$y[m] = \sum_l h_l[m] x[m - l],$$

where $h_l[m] := \sum_i a_i^b(m/W) \text{sinc}[l - \tau_i(m/W)W]$

Discrete - Time Baseband Model

- Discrete-time channel model

$$x[m] \rightarrow \boxed{h_l[m]} \rightarrow y[m] = \sum_l h_l[m]x[m - l]$$

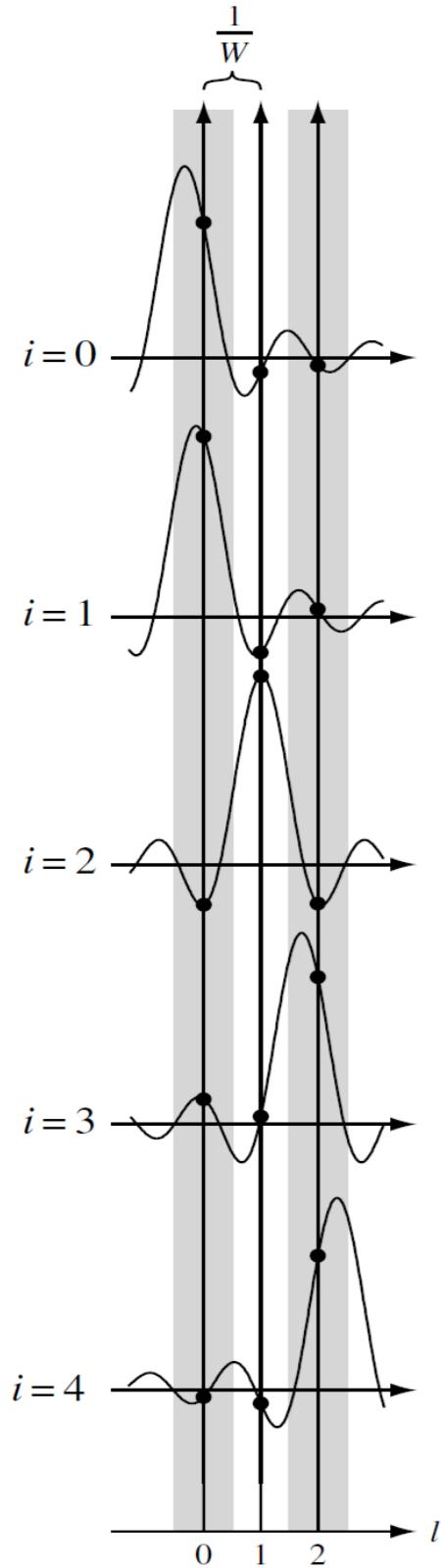
$$h_l[m] := \sum_i a_i^b(m/W) \text{sinc}[l - \tau_i(m/W)W]$$

- Note: the l -th tap h_l contains contributions mostly for the paths that have delays that lie inside the bin (roughly)

$$\left[\frac{l}{W} - \frac{1}{2W}, \frac{l}{W} + \frac{1}{2W} \right]$$

- System resolves the multipaths up to delays of $\frac{1}{W}$

Multipath Resolution



$$h_l[m] := \sum_i a_i^b(m/W) \text{sinc}[l - \tau_i(m/W)W]$$

- $\text{sinc}(t)$ vanish quickly outside of the interval $[-0.5, 0.5]$ (roughly)
- The peak of the i -th translated sinc lies at τ_i
- To contribute significantly to h_l , the delay must fall inside

$$\left[\frac{l}{W} - \frac{1}{2W}, \frac{l}{W} + \frac{1}{2W} \right]$$

Time and Frequency Coherence

Varying Channel Tap

- The discrete-time baseband channel model is the equivalent one in designing communication systems
- It only matters how the taps $h_l[m]$ vary over time m and carrier frequency f_c
- l -th tap of the discrete-time baseband channel model

$$\begin{aligned} h_l[m] &:= \sum_i a_i^b(m/W) \text{sinc}[l - \tau_i(m/W)W] \\ &= \sum_i a_i(t_m) e^{-j2\pi f_c \tau_i(t_m)} \text{sinc}[l - \tau_i(t_m)W] \quad t_m := \frac{m}{W} \\ &\approx \sum_{i \in l\text{-th delay bin}} a_i(t_m) e^{-j2\pi f_c \tau_i(t_m)} \end{aligned}$$

Difference in phases (over the paths that contribute significantly to the tap), causes variation of the tap gain

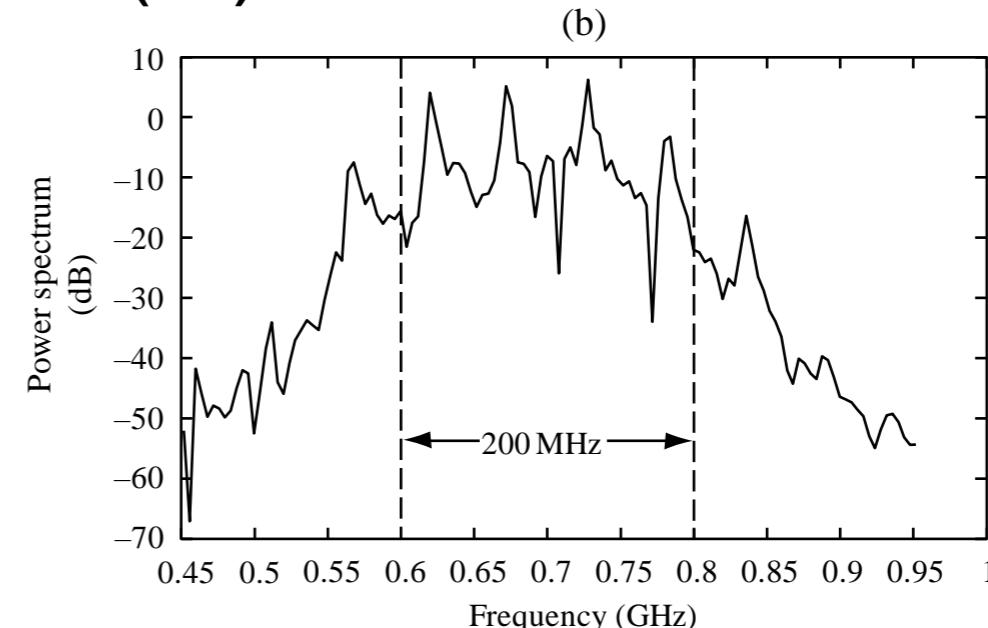
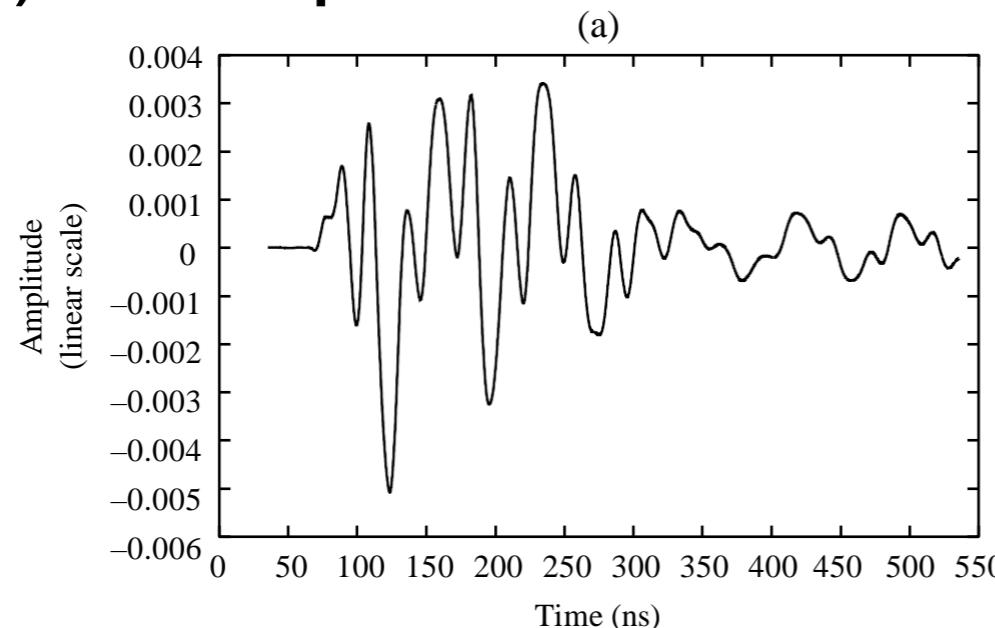
Frequency Variation

$$h_l[m] \approx \sum_{i \in l\text{-th delay bin}} a_i(t_m) e^{-j2\pi f_c \tau_i(t_m)}$$

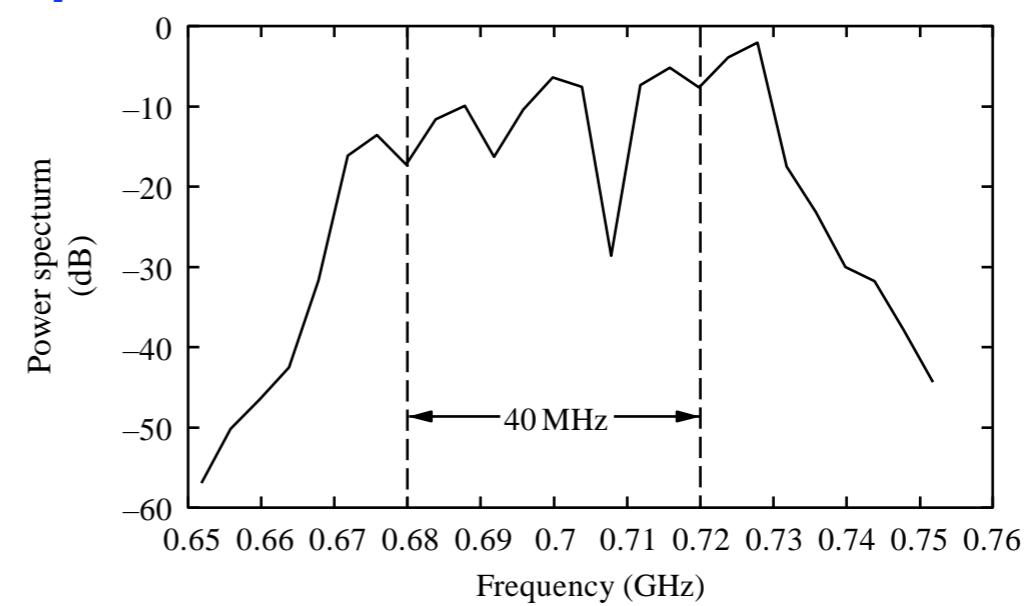
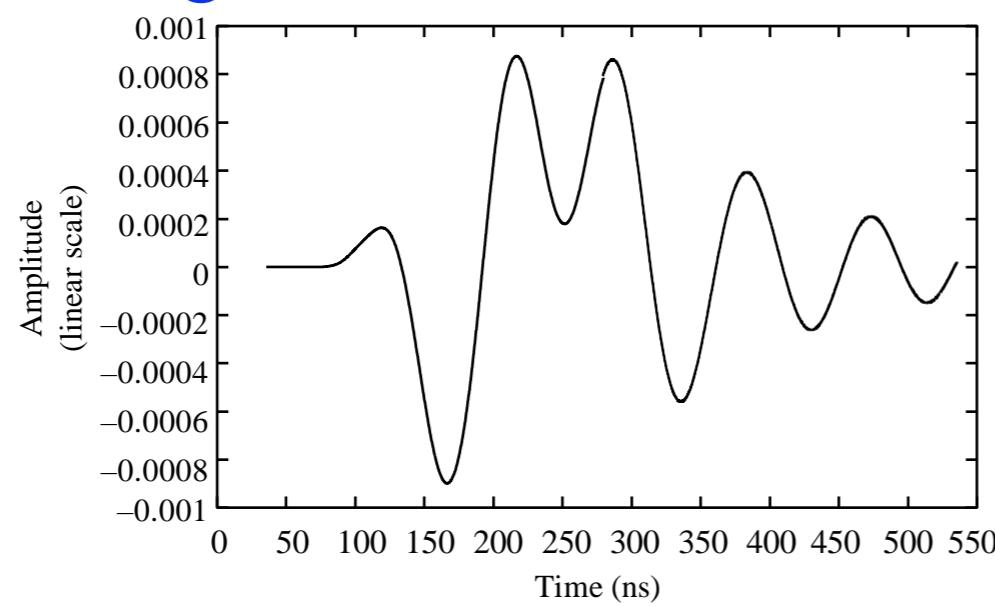
- Delay Spread $T_d := \max_{i,j} |\tau_i(t) - \tau_j(t)|$
- Coherence Bandwidth $W_c := \frac{1}{T_d}$
- For a system with bandwidth W
 - $W_c \gg W \implies$ single tap, flat fading
 - $W_c < W \implies$ multiple taps, frequency-selective fading

Flat and Frequency - Selective Fading

- Effective channel depends on both physical environment (W_c) and operation bandwidth (W)



Larger bandwidth, more paths can be resolved



Time Variation

$$h_l[m] \approx \sum_{i \in l\text{-th delay bin}} a_i(t_m) e^{-j2\pi f_c \tau_i(t_m)}$$

- Doppler Spread $D_s := \max_{i,j} f_c |\tau_i'(t) - \tau_j'(t)|$
- Coherence Time $T_c := \frac{1}{D_s}$
- For a system with delay requirement (application dependent) T

$T_c \gg T \implies$ slow fading

$T_c < T \implies$ fast fading

Representitive Numbers

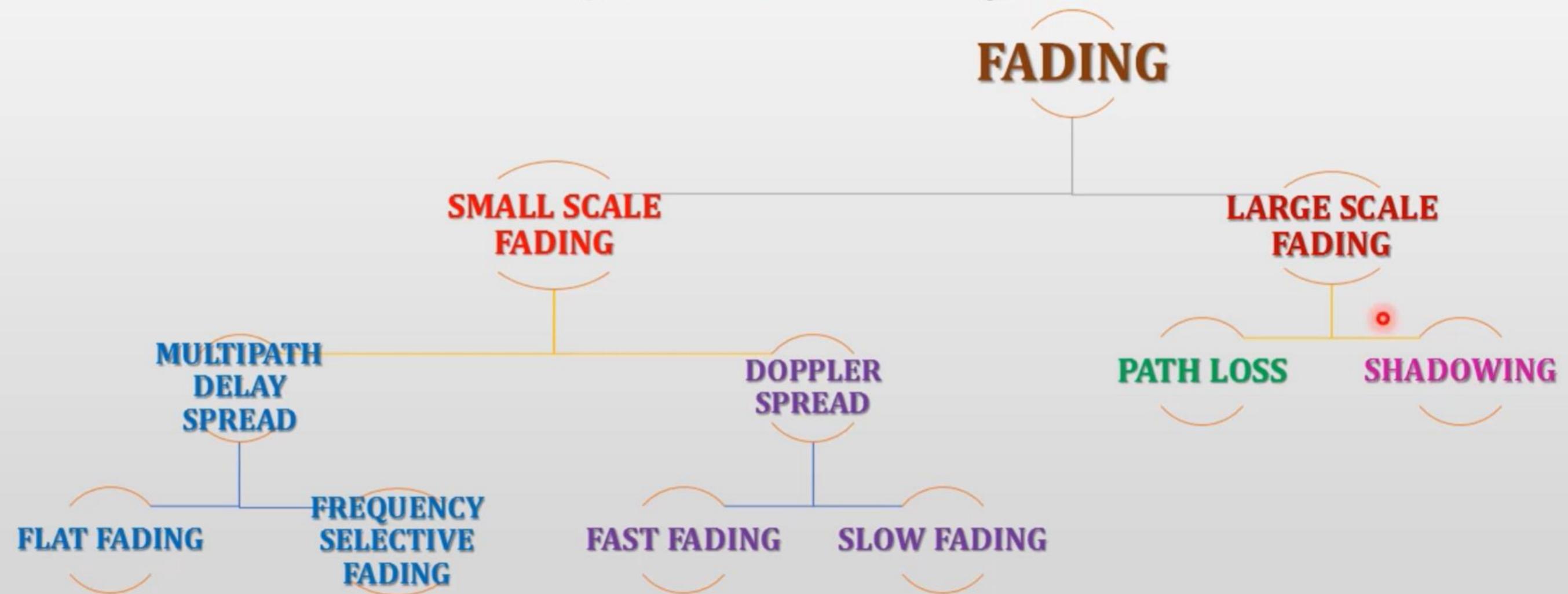
| Key channel parameters and time-scales | Symbol | Representative values |
|---|------------------|-----------------------|
| Carrier frequency | f_c | 1 GHz |
| Communication bandwidth | W | 1 MHz |
| Distance between transmitter and receiver | d | 1 km |
| Velocity of mobile | v | 64 km/h |
| Doppler shift for a path | $D = f_c v / c$ | 50 Hz |
| Doppler spread of paths corresponding to a tap | D_s | 100 Hz |
| Time-scale for change of path amplitude | d/v | 1 minute |
| Time-scale for change of path phase | $1/(4D)$ | 5 ms |
| Time-scale for a path to move over a tap | $c/(vW)$ | 20 s |
| Coherence time | $T_c = 1/(4D_s)$ | 2.5 ms |
| Delay spread | T_d | 1 μ s |
| Coherence bandwidth | $W_c = 1/(2T_d)$ | 500 kHz |

Types of Channels

| Types of channel | Defining characteristic |
|----------------------------|-----------------------------|
| Fast fading | $T_c \ll$ delay requirement |
| Slow fading | $T_c \gg$ delay requirement |
| Flat fading | $W \ll W_c$ |
| Frequency-selective fading | $W \gg W_c$ |
| Underspread | $T_d \ll T_c$ |

- Typical channels are underspread
- Coherence time T_c depends on carrier frequency and mobile speed, of the order of ms or more
- Delay spread T_d depends on distance to scatters and cell size, of the order of ns (indoor) to μ s (outdoor)

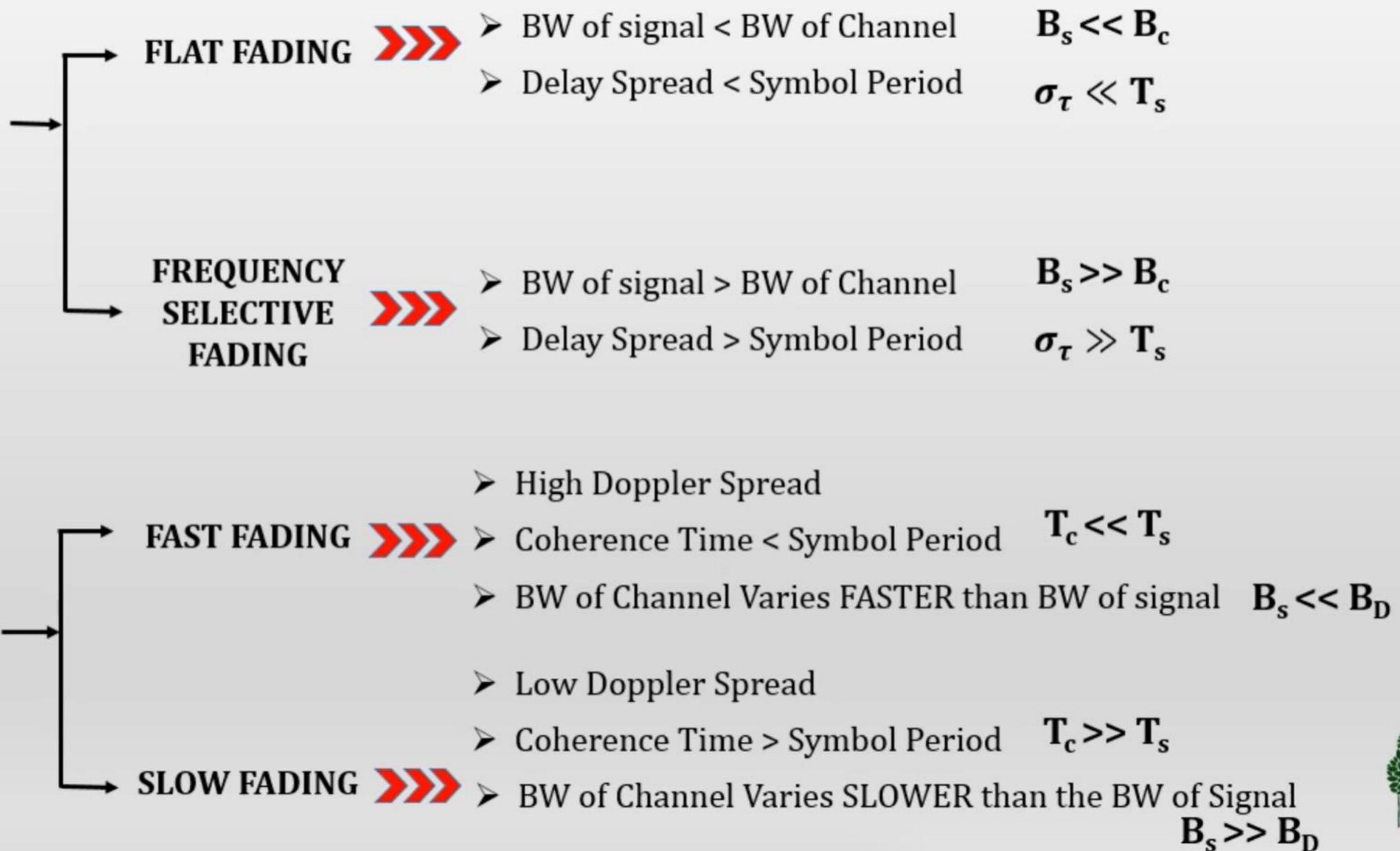
Types of Fading

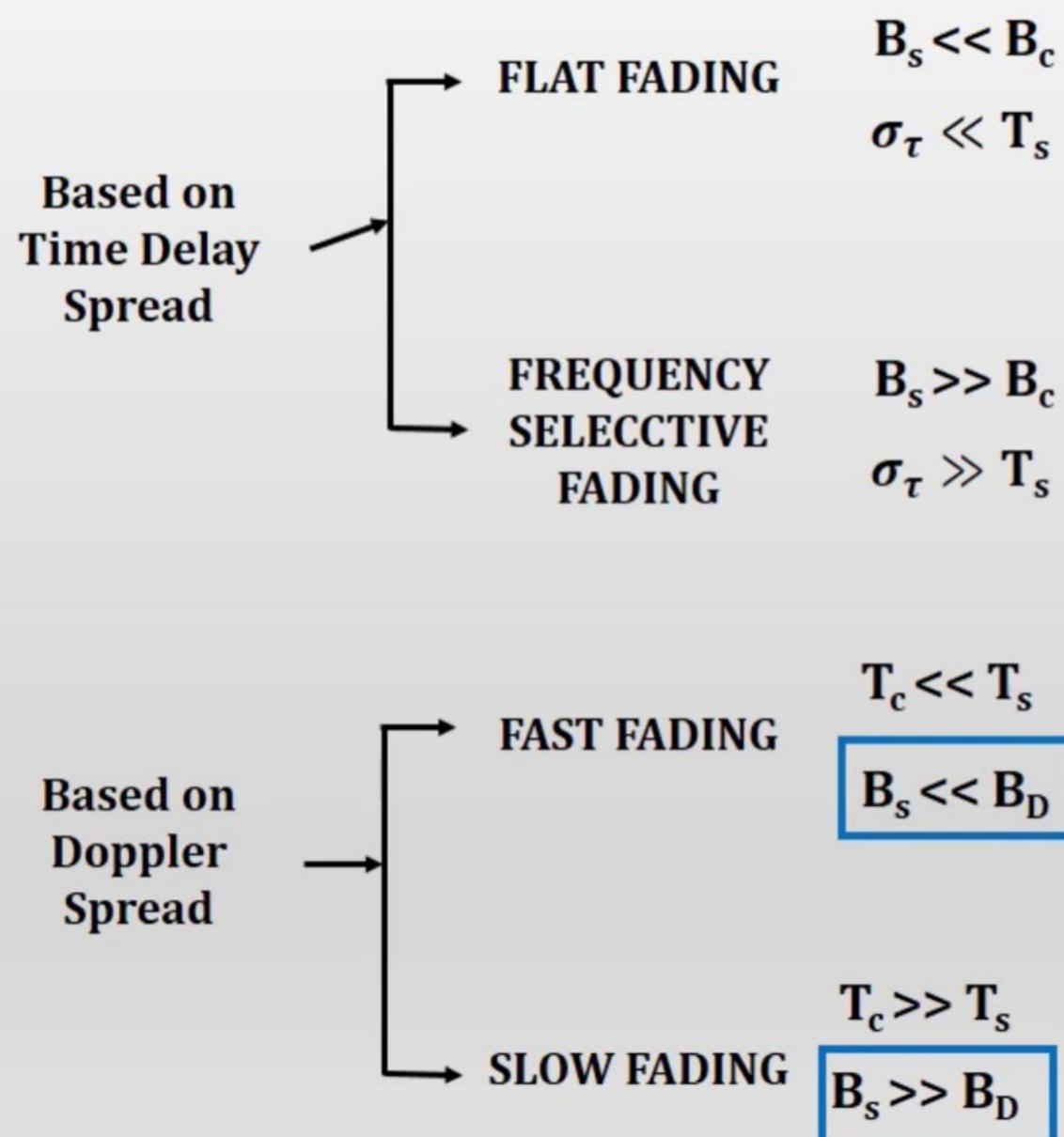


Types of Small Scale Fading

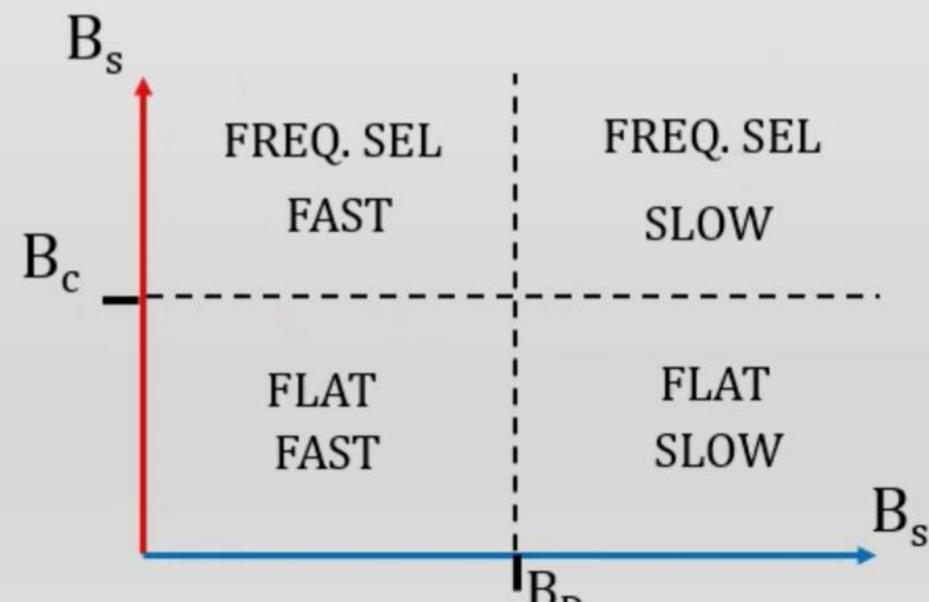
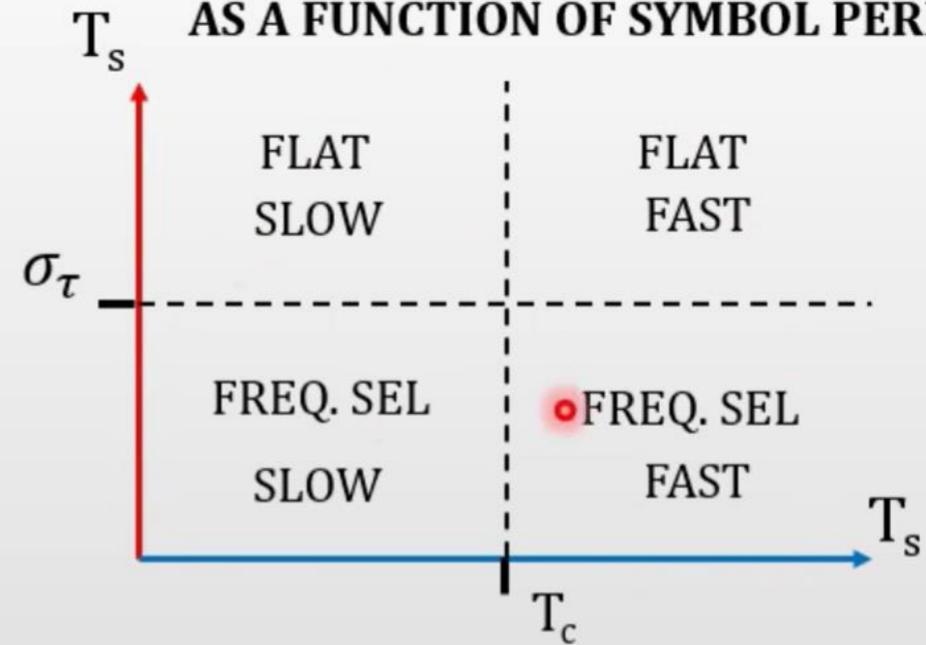
Based on
Time Delay
Spread

Leads to
TIME
Dispersion





MATRIX REPRESENTATION OF FADING AS A FUNCTION OF SYMBOL PERIOD

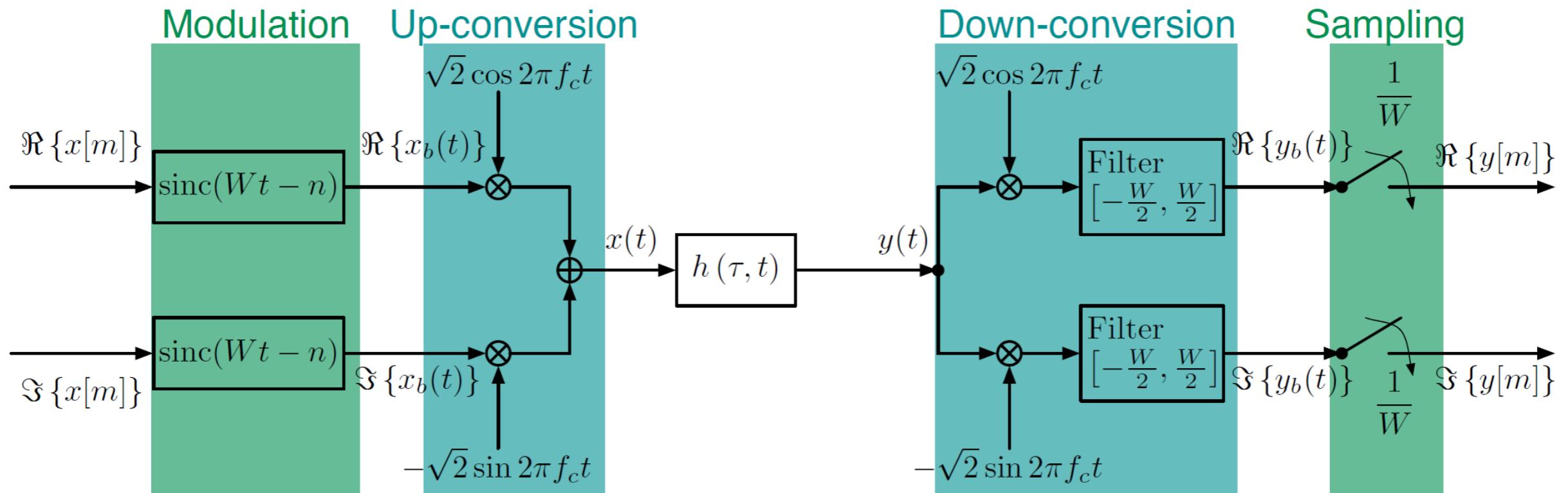


MATRIX REPRESENTATION OF FADING AS A FUNCTION OF SIGNAL BW

Stochastic Models

Recap: Deterministic Modeling

- Continuous-time Passband → Discrete-time Baseband:



- Continuous-time Passband (*Real*)

$$y(t) = \sum_i a_i(t) x(t - \tau_i(t)) \quad a_i(t): \text{gain of path } i \quad \tau_i(t): \text{delay of path } i$$

- Continuous-time Baseband (*Complex*)

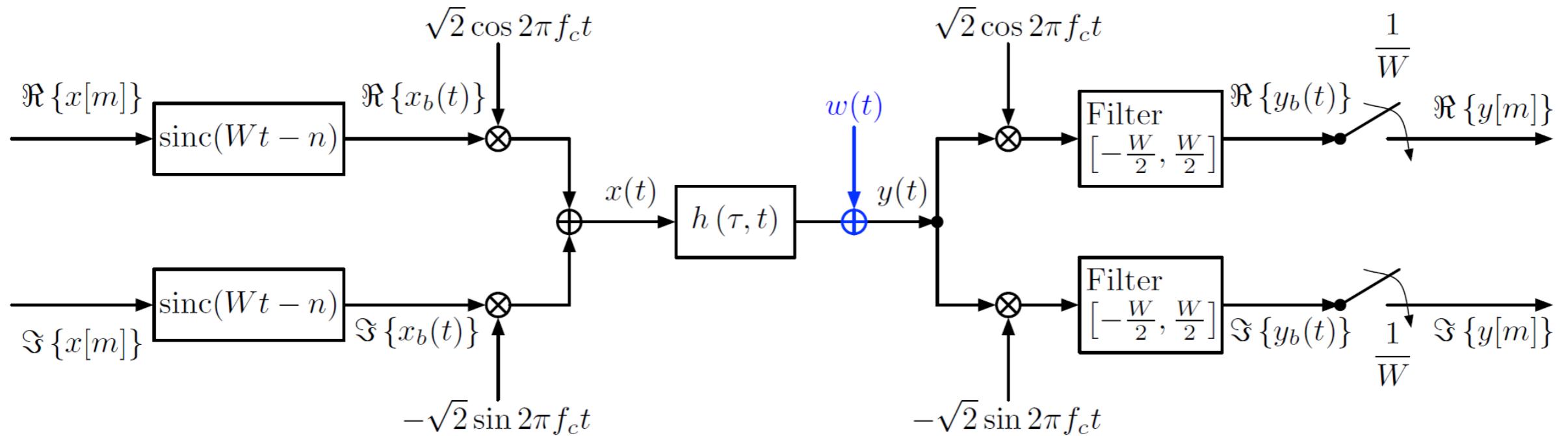
$$y_b(t) = \sum_i a_i^b(t) x_b(t - \tau_i(t)), \quad a_i^b(t) := a_i(t) e^{-j2\pi f_c \tau_i(t)}$$

- Discrete-time Baseband (*Complex*)

$$y[m] = \sum_l h_l[m] x[m - l], \quad h_l[m] := \sum_i a_i^b(m/W) \text{sinc}[l - \tau_i(m/W)W]$$

Noise

- Thermal noise at the receiver:



- Continuous-time Passband (*Real*)

$$y(t) = \sum_i a_i(t) x(t - \tau_i(t)) + w(t) \quad a_i(t): \text{gain of path } i \quad \tau_i(t): \text{delay of path } i$$

- Continuous-time Baseband (*Complex*)

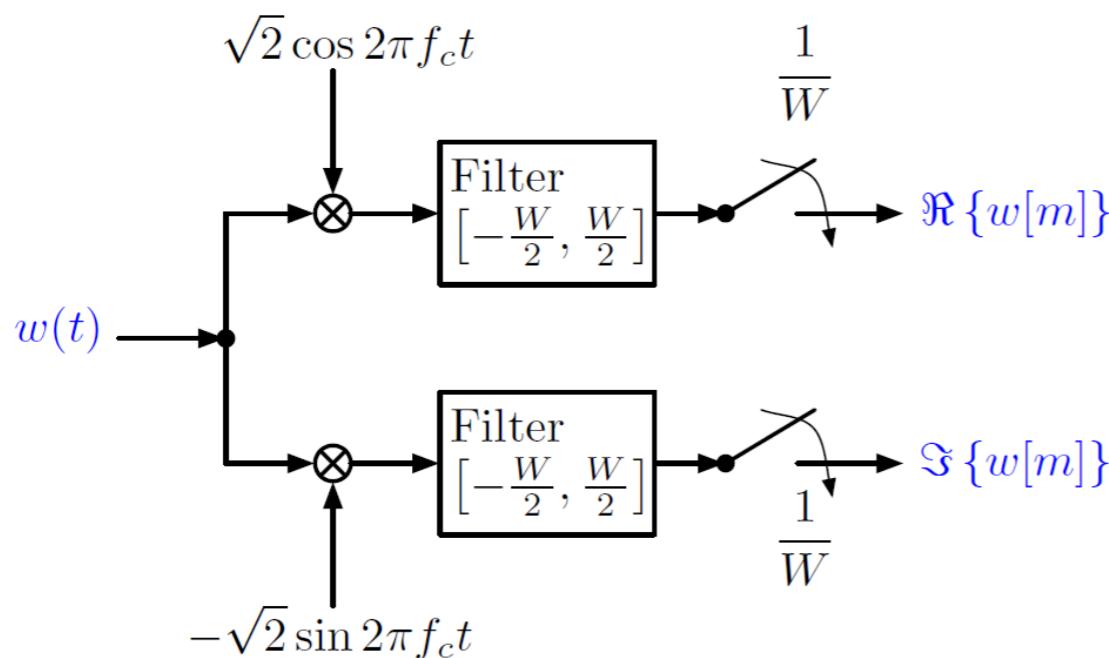
$$y_b(t) = \sum_i a_i^b(t) x_b(t - \tau_i(t)) + w_b(t), \quad a_i^b(t) := a_i(t) e^{-j2\pi f_c \tau_i(t)}$$

- Discrete-time Baseband (*Complex*)

$$y[m] = \sum_l h_l[m] x[m - l] + w[m], \quad h_l[m] := \sum_i a_i^b(m/W) \text{sinc}[l - \tau_i(m/W)W]$$

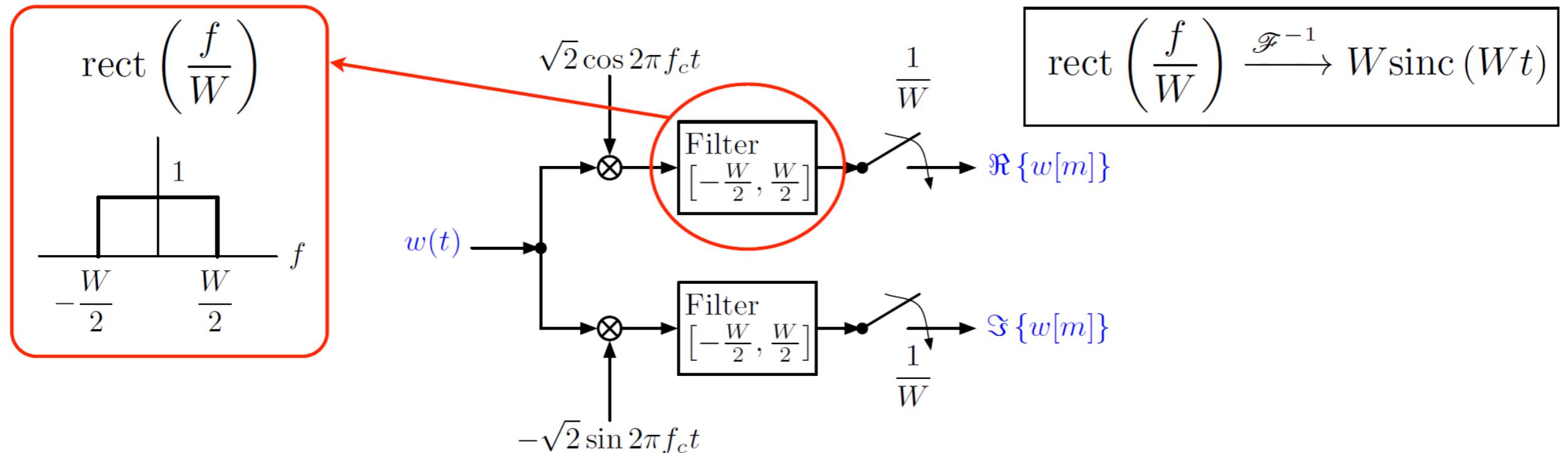
Additive White Noise Model

- Additive White Gaussian Noise (AWGN)
 - Standard modeling for thermal noise
 - $\{w(t)\}$: zero-mean (real) white Gaussian process with spectral density $N_0/2$
 - In other words, $\mathbb{E}[w(0)w(t)] = \frac{N_0}{2}\delta(t)$
- Discrete-time baseband equivalent noise:



- System is linear \rightarrow can separate the noise out
- Rectangle filter in frequency \Leftrightarrow sinc in time

Equivalent Discrete-Time Baseband Noise



$$\begin{aligned}\Re \{w[m]\} &= \int_{-\infty}^{\infty} w(t) \underbrace{\left[\sqrt{2}W \cos(2\pi f_c t) \text{sinc}(Wt - m) \right]}_{\psi_{m,1}(t)} dt \\ &= \boxed{\langle w(t), \psi_{m,1}(t) \rangle}\end{aligned}$$

$$\begin{aligned}\Im \{w[m]\} &= \int_{-\infty}^{\infty} w(t) \underbrace{\left[-\sqrt{2}W \sin(2\pi f_c t) \text{sinc}(Wt - m) \right]}_{\psi_{m,2}(t)} dt \\ &= \boxed{\langle w(t), \psi_{m,2}(t) \rangle}\end{aligned}$$

Equivalent Discrete-Time Baseband Noise

$$\Re \{w[m]\} = \int_{-\infty}^{\infty} w(t) \underbrace{\left[\sqrt{2W} \cos(2\pi f_c t) \operatorname{sinc}(Wt - m) \right]}_{\psi_{m,1}(t)} dt$$

$$= \boxed{\langle w(t), \psi_{m,1}(t) \rangle}$$

$$\Im \{w[m]\} = \int_{-\infty}^{\infty} w(t) \underbrace{\left[-\sqrt{2W} \sin(2\pi f_c t) \operatorname{sinc}(Wt - m) \right]}_{\psi_{m,2}(t)} dt$$

$$= \boxed{\langle w(t), \psi_{m,2}(t) \rangle}$$

Fact.

$\{\psi_{m,1}(t), \psi_{m,2}(t) \mid m \in \mathbb{Z}\}$ forms an orthogonal set of waveforms.

- The real and the imaginary parts of $w[m]$ are
 - Both Gaussian with zero-mean and variance $WN_0/2$
 - Independent and identically distributed (i.i.d.) over time (m)
 - “White” discrete-time processes

Circular Symmetric Complex Gaussian

$$w[m] = \Re\{w[m]\} + j\Im\{w[m]\}$$

$$\Re\{w[m]\} \sim \mathcal{N}\left(0, \frac{WN_0}{2}\right), \quad \Im\{w[m]\} \sim \mathcal{N}\left(0, \frac{WN_0}{2}\right)$$

$(\Re\{w[m]\}, \Im\{w[m]\})$: independent

$\iff w[m] \sim \mathcal{CN}(0, WN_0)$: circular symmetric complex Gaussian

Fact.

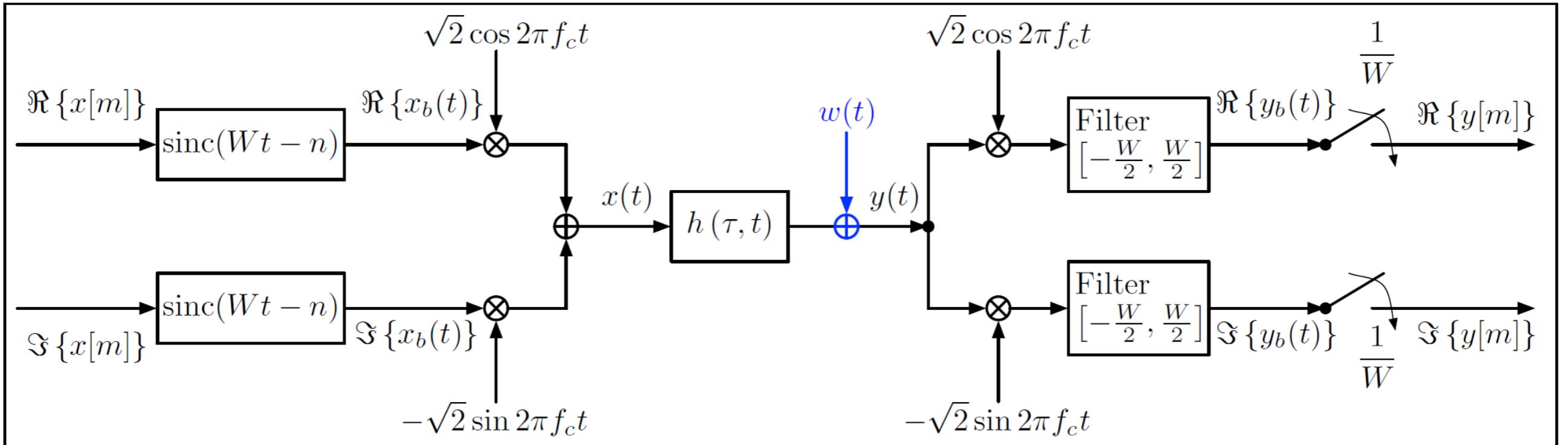
X is circular symmetric if $e^{i\theta}X \stackrel{\text{d}}{=} X$ for all θ the same distribution

Fact.

$\{w[m] \mid m \in \mathbb{Z}\}$ is a zero-mean white circular symmetric complex Gaussian process with auto-correlation function

$$R[m] := \mathbb{E}[w[n+m]w[n]^*] = WN_0\delta[m]$$

Discrete Baseband Model with Noise



|||

$$x[m] \rightarrow h_l[m] \rightarrow y[m] = \sum_l h_l[m] x[m-l] + w[m]$$

$$w[m] \sim \mathcal{CN}(0, WN_0),$$

where $\frac{N_0}{2}$ is the spectral density of white Gaussian process $\{w(t)\}$

Fading

$$y[m] = \sum_l h_l[m]x[m - l] + w[m],$$

$$\text{where } h_l[m] := \sum_i a_i^b(m/W) \text{sinc}[l - \tau_i(m/W)W]$$
$$w[m] \sim \mathcal{CN}(0, WN_0)$$

- Additive noise $w[m]$
 - Essentially completely random, no correlation over time
 - Largely depends on nature
 - Can be dealt with using digital (wired) communication techniques
- Filter taps $h_l[m]$
 - Varying over time and frequency
 - Largely depends on nature
 - Why not use stochastic models for taps as well?

Rayleigh Fading

- Many small scattered paths for each tap:
 - Phase for each path is uniformly distributed over $[0, 2\pi]$
$$h_l[m] \approx \sum_i a_i(t_m) e^{-j2\pi f_c \tau_i(t_m)}$$
 - For each path it is a circular symmetric random variable
- Each tap: sum of many small independent *circular symmetric* random variables
 - By Central Limit Theorem (CLT), we can model
$$h_l[m] \sim \mathcal{CN}(0, \sigma_l^2)$$
 - Zero-mean because of rich scattering

Rician Fading

- If there is a strong line-of-sight path, then model it as

$$h_l[m] = \sqrt{\frac{\kappa}{\kappa + 1}} \sigma_l e^{j\theta} + \sqrt{\frac{1}{\kappa + 1}} \mathcal{CN}(0, \sigma_l^2)$$

Line-of-sight Scattered Multipath

- K-Factor κ :
 - The larger it is, the more deterministic the channel will be.