Digital Communications Chapter 13 Fading Channels I: Characterization and **Signaling**

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[13.1 Characterization of fading](#page-1-0) [multipath channels](#page-1-0)

The multipath fading channels with additive noise

Time spread phenomenon of multipath channels (Unpredictable) Time-variant factors

- **•** Delay
- Number of spreads
- Size of the receive pulses

Transmitted signal

$$
s(t) = \text{Re}\left\{s_{\ell}(t)e^{i2\pi f_c t}\right\}
$$

Received signal in absence of additive noise

$$
r(t) = \int_{-\infty}^{\infty} c(\tau; t) s(t - \tau) d\tau
$$

\n
$$
= \int_{-\infty}^{\infty} c(\tau; t) \text{Re} \{ s_{\ell}(t - \tau) e^{i2\pi f_c(t - \tau)} \} d\tau
$$

\n
$$
= \text{Re} \{ \left(\int_{-\infty}^{\infty} c(\tau; t) e^{-i2\pi f_c \tau} s_{\ell}(t - \tau) d\tau \right) e^{i2\pi f_c t} \}
$$

\n
$$
= \text{Re} \{ \left(s_{\ell}(t) \star \underbrace{c(\tau; t) e^{-i2\pi f_c \tau}}_{\pm c_{\ell}(\tau; t)} \right) e^{i2\pi f_c t} \}
$$

In Slide 2-28, we define the lowpass equivalent system as

$$
\begin{cases}\nX_{\ell}(f) \triangleq 2X_{+}(f + f_{0}) \\
Y_{\ell}(f) \triangleq 2Y_{+}(f + f_{0}) \\
H_{\ell}(f) \triangleq 2H_{+}(f + f_{0})\n\end{cases} \text{ and obtain } \begin{cases}\nY_{\ell}(f) = \frac{1}{2}X_{\ell}(f)H_{\ell}(f) \\
(i.e., y_{\ell}(f) = \frac{1}{2}x_{\ell}(t) \star h_{\ell}(t))\n\end{cases}
$$

Here, under a time-invariant $c(\tau;t)$ = $c(\tau)$, we actually define

$$
c_{\ell}(\tau) \triangleq c(\tau) e^{-i 2\pi f_c \tau},
$$

equivalently,

$$
C_{\ell}(f)=\int_{-\infty}^{\infty}c(\tau)e^{-i2\pi f_c\tau}e^{-i2\pi f\tau}d\tau=C(f+f_0).
$$

Thus, the new "lowpass equivalence" yields

$$
\begin{cases}\nS_{\ell}(f) \triangleq 2S_{+}(f + f_{0}) \\
R_{\ell}(f) \triangleq 2R_{+}(f + f_{0}) \\
C_{\ell}(f) \triangleq C(f + f_{0})\n\end{cases}\n\Rightarrow\n\begin{cases}\nR_{\ell}(f) = S_{\ell}(f)C_{\ell}(f) \\
(i.e., r_{\ell}(f) = s_{\ell}(t) \times c_{\ell}(t))\n\end{cases}
$$

An advantage of this new equivalence is that the statistics of $c(\tau; t) =$ $|c_{\ell}(\tau ;t)|$ can be determined from the statistics of $c_{\ell}(\tau ;t).$

Note that for a time-varying system, t and τ specifically denote time argument and convolution argument, respectively!

We should perhaps write $s_{\ell}(t) \star c_{\ell}(\tau)$ and $s_{\ell}(t) \star c_{\ell}(\tau; t)$, which respectively denote:

$$
s_{\ell}(t)\star c_{\ell}(\tau)=\int_{-\infty}^{\infty}c_{\ell}(\tau)s_{\ell}(t-\tau)d\tau
$$

and

$$
s_{\ell}(t)\star c_{\ell}(\tau;t)=\int_{-\infty}^{\infty}c_{\ell}(\tau;t)s_{\ell}(t-\tau)d\tau.
$$

From the previous slide, we know

$$
c_{\ell}(\tau;t) = c(\tau;t)e^{-i2\pi f_c \tau} \text{ and } c(\tau;t) = |c_{\ell}(\tau;t)|.
$$

Rayleigh and Rician

Measurements suggest that in certain environment, $c(\tau; t) = |c_{\ell}(\tau; t)| \ge 0$ can be Rayleigh distributed or Rician distributed. As a consequence, such $c(\tau; t)$ can be modeled by letting $c_{\ell}(\tau; t)$ be a 2-D Gaussain random process in t (not in τ).

- **If** $c_{\ell}(\tau; t)$ zero mean, $c(\tau; t) = |c_{\ell}(\tau; t)|$ is Rayleigh distributed. The channel $c(\tau;t)$ is said to be a Rayleigh fading channel.
- **If** $c_{\ell}(\tau; t)$ nonzero mean, $c(\tau; t) = |c_{\ell}(\tau; t)|$ is Rician distributed. The channel $c(\tau;t)$ is said to be a Rician fading channel.

When diversity technique is used, $c(\tau; t) = |c_{\ell}(\tau; t)|$ is well modeled by Nakagami m-distribution.

[13.1-1 Channel correlation](#page-8-0) [functions and power spectra](#page-8-0)

Assumption (WSS)

 $c_{\ell}(\tau; t)$ is WSS in t.

$$
R_{c_{\ell}}(\bar{\tau},\tau;\Delta t) = \mathbb{E}\left\{c_{\ell}(\bar{\tau};t+\Delta t)c_{\ell}^{*}(\tau;t)\right\}
$$

is only a function of time difference Δt .

Assumption (Uncorrelated scattering or US of a WSS channel) For $\bar{\tau}$ ≠ τ , $c_{\ell}(\bar{\tau};t_1)$ and $c_{\ell}(\tau;t_2)$ are uncorrelated for any t_1,t_2 .

 \bullet τ is the convolution argument and actually represents the delay for a certain path.

Assumption (Math definition of US)

$$
R_{c_{\ell}}(\bar{\tau},\tau;\Delta t) = R_{c_{\ell}}(\tau;\Delta t)\delta(\bar{\tau}-\tau)
$$

Multipath intensity profile of a WSSUS channel

• The multipath intensity profile or delay power spectrum for a WSSUS multipath fading channel is given by:

$$
R_{c_{\ell}}(\tau) = R_{c_{\ell}}(\tau; \Delta t = 0).
$$

• It can be interpreted as the average power output of the channel as a function of the path delay τ .

$$
\mathbb{E}[|r_{\ell}(t)|^{2}] = \mathbb{E}\Bigg[\int_{-\infty}^{\infty} c_{\ell}(\bar{\tau}; t) s_{\ell}(t - \bar{\tau}) d\bar{\tau} \int_{-\infty}^{\infty} c_{\ell}^{*}(\tau; t) s_{\ell}^{*}(t - \tau) d\tau\Bigg]
$$
\n
$$
= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \mathbb{E}\Big[c_{\ell}(\bar{\tau}; t) c_{\ell}^{*}(\tau; t)\Big] \mathbb{E}[s_{\ell}(t - \bar{\tau}) s_{\ell}^{*}(t - \tau)] d\bar{\tau} d\tau
$$
\n
$$
= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{c_{\ell}}(\tau; 0) \delta(\bar{\tau} - \tau) \mathbb{E}[s_{\ell}(t - \bar{\tau}) s_{\ell}^{*}(t - \tau)] d\bar{\tau} d\tau
$$
\n
$$
= \int_{-\infty}^{\infty} R_{c_{\ell}}(\tau; 0) \mathbb{E}[|s_{\ell}(t - \tau)|^{2}] d\tau = \int_{-\infty}^{\infty} R_{c_{\ell}}(\tau) \mathbb{E}[|s_{\ell}(t - \tau)|^{2}] d\tau
$$

Multipath spread of a WSSUS channel

- **multipath spread or delay spread of a WSSUS** multipath fading channel
	- **multipath spread** is the range of τ over which $R_{c_{\ell}}(\tau)$ is
essentially non-range it is usually denoted by τ essentially non-zero; it is usually denoted by T_m .

- Each τ corresponds to one path.
- No Tx power will essentially remain at Rx for paths with delay $\tau > T_m$.

The transfer function of a channel impulse response $c_{\ell}(\tau; t)$ is the Fourier transform with respect to the convolutional argument τ :

$$
\mathbf{C}_{\ell}(f;t) = \int_{-\infty}^{\infty} c_{\ell}(\tau;t) e^{-i 2\pi f \tau} d\tau
$$

Property: If $c_{\ell}(\tau; t)$ is WSS, so is $\mathbf{C}_{\ell}(f; t)$.

The autocorrelation function of WSS $\mathbf{C}_\ell(f; t)$ is equal to:

$$
R_{\mathbf{C}_{\ell}}(\bar{f},f;\Delta t) = \mathbb{E}\left\{ \mathbf{C}_{\ell}(\bar{f};t+\Delta t)\mathbf{C}_{\ell}^{*}(f;t)\right\}
$$

With an additional US assumption,

$$
R_{\mathbf{C}_{\ell}}(\bar{f}, f; \Delta t)
$$
\n
$$
= \mathbb{E}\left\{\mathbf{C}_{\ell}(\bar{f}; t + \Delta t)\mathbf{C}_{\ell}^{*}(f; t)\right\}
$$
\n
$$
= \mathbb{E}\left\{\int_{-\infty}^{\infty} c_{\ell}(\bar{\tau}; t + \Delta t)e^{-i2\pi\bar{f}\bar{\tau}}d\bar{\tau}\int_{-\infty}^{\infty} c_{\ell}^{*}(\tau; t)e^{i2\pi f\tau}d\tau\right\}
$$
\n
$$
= \int_{-\infty}^{\infty}\int_{-\infty}^{\infty} R_{c_{\ell}}(\tau; \Delta t)\delta(\bar{\tau} - \tau)e^{i2\pi(f\tau - \bar{f}\bar{\tau})}d\tau d\bar{\tau}
$$
\n
$$
= \int_{-\infty}^{\infty} R_{c_{\ell}}(\tau; \Delta t)e^{-i2\pi(\bar{f}-f)\tau}d\tau
$$
\n
$$
= R_{\mathbf{C}_{\ell}}(\Delta f; \Delta t), \text{ where } \Delta f = \bar{f} - f.
$$

For a WSSUS multipath fading channel,

$$
R_{\mathbf{C}_{\ell}}(\Delta f; \Delta t) = \mathbb{E}\left\{ \mathbf{C}_{\ell}(f + \Delta f; t + \Delta t)\mathbf{C}_{\ell}^{*}(f; t) \right\}
$$

This is often called spaced-frequency, spaced-time correlation function of a WSSUS channel.

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Note that $R_{\ell}(f) \neq S_{\ell}(f)C_{\ell}(f; t)$, where

$$
R_{\ell}(f)=\int_{-\infty}^{\infty}r_{\ell}(t)e^{-i2\pi ft}dt \text{ and } S_{\ell}(f)=\int_{-\infty}^{\infty}s_{\ell}(t)e^{-i2\pi ft}dt.
$$

We only have

$$
r_{\ell}(t) = \int_{-\infty}^{\infty} c_{\ell}(\tau; t) s_{\ell}(t - \tau) d\tau
$$

\n
$$
= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} \mathbf{C}_{\ell}(f; t) e^{i 2\pi f \tau} d\tau \right) s_{\ell}(t - \tau) d\tau
$$

\n
$$
= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} s_{\ell}(t - \tau) e^{i 2\pi f \tau} d\tau \right) \mathbf{C}_{\ell}(f; t) d\tau
$$

\n
$$
= \int_{-\infty}^{\infty} \left(\int_{-\infty}^{\infty} s_{\ell}(u) e^{i 2\pi f (t - u)} d\tau \right) \mathbf{C}_{\ell}(f; t) d\tau \quad (\text{Let } u = t - \tau)
$$

\n
$$
= \int_{-\infty}^{\infty} S_{\ell}(f) \mathbf{C}_{\ell}(f; t) e^{i 2\pi f t} d\tau
$$

Coherent bandwidth

Summarize from the last equality of the previous derivation (in red): $R_{\mathsf{C}_{\ell}}\left(\Delta f ; \Delta t\right) = \int_{-\infty}^{\infty} R_{c_{\ell}}\left(\tau ; \Delta t\right) e^{-i 2\pi (\Delta f) \tau} d\tau$

For the case of $\Delta t = 0$, we have

$$
\underbrace{R_{\mathbf{C}_{\ell}}(\Delta f)}_{\text{spaced-frequency}} = \int_{-\infty}^{\infty} R_{c_{\ell}}(\tau) e^{-i 2\pi (\Delta f) \tau} d\tau
$$

- Recall that $R_{c_\ell}(\tau) = 0$ outside $[0, T_m)$.
- $(\Delta f)_c$ = 1 $\frac{1}{\mathcal{T}_m}$ is called coherent bandwidth.
- From Slide 13-11, $\mathbb{E}[|r_{\ell}(t)|^2] = \int$ $\int_{-\infty}^{\infty} R_{c_{\ell}}(\tau) \mathbb{E}[|s_{\ell}(t-\tau)|^2] d\tau$ −∞

$$
\Rightarrow \int_{-\infty}^{\infty} \mathbb{E}[|r_{\ell}(t)|^{2}] e^{-i 2\pi (\Delta f)t} dt
$$

= $R_{\mathsf{C}_{\ell}}(\Delta f) \int_{-\infty}^{\infty} \mathbb{E}[|s_{\ell}(t)|^{2}] e^{-i 2\pi (\Delta f)t} dt$

FIGURE 13.1-3 Relationship between $R_C(\Delta f)$ and $R_c(\tau)$.

Example.

$$
R_{\mathbf{C}_{\ell}}(\bar{f},f;0)=\mathbb{E}\left\{\mathbf{C}_{\ell}(\bar{f};t)\mathbf{C}_{\ell}^{*}(f;t)\right\}\text{ and }r_{\ell}(t)=\int_{-\infty}^{\infty}S_{\ell}(f)\mathbf{C}_{\ell}(f;t)e^{i2\pi ft}df
$$

If $\bar{f} - f > (\Delta f)_c$, $R_{\mathbf{C}_{\ell}}(\bar{f}, f; 0)$ will be essentially small (nearly uncorrelated or nearly independent if Gaussian).

Thus, two sinusoids $S_{\ell}(\bar{f})$ and $S_{\ell}(f)$ with frequency separation greater than $(\Delta f)_c$ are respectively multiplied by nearly independent $\mathbf{C}_{\ell}(\vec{f};t)$ and $\mathbf{C}_{\ell}(f;t)$ and hence are affected very differently by the channel.

If signal transmitted bandwidth $B<(\Delta f)_c$, the channel is called frequency non-selective.

- For frequency selective channels, the signal shape is more severely distorted than that of frequency non-selective channels.
- Criterion for frequency selectivity:

$$
B > (\Delta f)_c \quad \Leftrightarrow \quad \frac{1}{T} > \frac{1}{T_m} \quad \Leftrightarrow \quad T < T_m.
$$

Time varying characterization: Doppler

Doppler effect appears via the argument Δt .

Doppler power spectrum of a WSSUS channel

The Doppler power spectrum is

$$
S_{\mathbf{C}_{\ell}}(\lambda)=\int_{-\infty}^{\infty}R_{\mathbf{C}_{\ell}}(\Delta f=0;\Delta t)e^{-i2\pi\lambda(\Delta t)}d(\Delta t),
$$

where λ is referred to as the **Doppler frequency**.

- B_d = Doppler spread is the range such that $\mathcal{S}_{\mathbf{C}_{\ell}}(\lambda)$ is
essentially zero essentially zero.
- $(\Delta t)_c = \frac{1}{B_c}$ $\frac{1}{B_d}$ is called the coherent time.
- If symbol period $T > (\Delta t)_{c}$, the channel is classified as
Fast Fading Fast Fading.
	- I.e., channel statistics changes within one symbol!
- If symbol period $T < (\Delta t)_c$, the channel is classified as
Slow Fading Slow Fading.

Operational Characteristic of $\mathcal{S}_{{\sf{C}}_\ell}(\lambda)$

We can obtain as similarly from Slide 13-11 that

$$
\bar{R}_{r_{\ell}}(\Delta t) = \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{T} \mathbb{E} [r_{\ell}(t + \Delta t) r_{\ell}^{*}(t)] dt
$$
\n
$$
= \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{T} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} R_{c_{\ell}}(\tau; \Delta t) \delta(\bar{\tau} - \tau)
$$
\n
$$
\times \mathbb{E} [s_{\ell}(t + \Delta t - \bar{\tau}) s_{\ell}^{*}(t - \tau)] d\bar{\tau} d\tau dt
$$
\n
$$
= \lim_{T \to \infty} \frac{1}{2T} \int_{-T}^{T} \int_{-\infty}^{\infty} R_{c_{\ell}}(\tau; \Delta t) \mathbb{E} [s_{\ell}(t + \Delta t - \tau) s_{\ell}^{*}(t - \tau)] d\tau dt
$$
\n
$$
= \int_{-\infty}^{\infty} R_{c_{\ell}}(\tau; \Delta t) \bar{R}_{s_{\ell}}(\Delta t) d\tau
$$
\n
$$
= R_{c_{\ell}}(\Delta f = 0; \Delta t) \bar{R}_{s_{\ell}}(\Delta t)
$$
\n
$$
\Rightarrow \bar{S}_{r_{\ell}}(\lambda) = S_{c_{\ell}}(\lambda) \times \bar{S}_{s_{\ell}}(\lambda).
$$

FIGURE 13.1-4 Relationship between $R_C(\Delta t)$ and $\mathcal{S}_C(\lambda)$.

Scattering function

Summary:

 $R_{c_{\ell}}(\tau;\Delta t)$ Channel autocorrelation function $1-D$ FT: $\left\{\right.$ $R_{\mathsf{C}_{\ell}}(\Delta f; \Delta t) = \mathcal{F}_{\tau} \{R_{c_{\ell}}(\tau; \Delta t)\}$ Spaced-freq spaced-time correlation func $S(\tau; \lambda) = \mathcal{F}_{\Delta t} \{ R_{c_\ell}(\tau; \Delta t) \}$ Scattering function 2D FT: $S_{\mathbf{C}_{\ell}}(\Delta f; \lambda) = \mathcal{F}_{\tau, \Delta t} \{R_{c_{\ell}}\}$ $(\tau; \Delta t)$ } Doppler power
spectrum ($\Delta f = 0$) $R_{\mathsf{C}_{\ell}}(\Delta f ; \Delta t)$ Spaced-freq spaced-time correlation function $1-D$ FT: $\left\{ \right.$ $R_{c_{\ell}}(\tau; \Delta t) = \mathcal{F}_{\Delta f}^{-1} \left\{ R_{\mathbf{C}_{\ell}}(\Delta f; \Delta t) \right\}$ Chan autocorr function $S_{\mathbf{C}_{\ell}}(\Delta f; \lambda) = \mathcal{F}_{\Delta t} \{ R_{\mathbf{C}_{\ell}}(\Delta f; \Delta t) \}$ Doppler power
spectrum (Δf) spectrum $(\Delta f = 0)$

2D FT: $S(\tau; \lambda) = \mathcal{F}_{\Delta f}^{-1} \mathcal{F}_{\Delta t}$ { $R_{\mathbf{C}_{\ell}}$ Scattering function

Scattering function

• The scattering function can be used to identify both "delay spread" and "Doppler spread."

Example. Medium-range tropospheric scatter channel

Scattering function of a medium-range tropospheric scatter channel. The taps delay increment is $0.1 \mu s$.

> \rightarrow B_d = Doppler spread, varies with paths $=$ (often) 3dB bandwidth \approx 1Hz \sim 30Hz

The **median delay spread** is the 50% value, meaning that 50% of all channels has a delay spread that is lower than the median value. Clearly, the median value is not so interesting for designing a wireless link, because you want to guarantee that the link works for at least 90% or 99% of all channels. Therefore the second column gives the measured **maximum** delay spread values. The reason to use maximum delay spread instead of a 90% or 99% value is that many papers only mention the maximum value. From the papers that do present cumulative distribution functions of their measured delay spreads, it can be deduced that the 99% value is only a few percent smaller than the maximum delay spread.

Measured delay spreads in frequency range of 800M to 1.5 GHz (surveyed by Richard van Nee, Lucent Technologies, Nov. 1997)

Measured delay spreads in frequency range of 1.8 GHz to 2.4 GHz (surveyed by Richard van Nee)

Measured delay spreads in frequency range of 4 GHz to 6 GHz (surveyed by Richard van Nee)

Conclusion by Richard van Nee: Measurements done at different frequencies show the multipath channel characteristics are almost the same from 1 to 5 GHz.

Jakes' model: Example 13.1-3

Jakes' model

A widely used model for Doppler power spectrum is the so-called Jakes' model (Jakes, 1974)

$$
R_{\mathbf{C}_{\ell}}(\Delta t) = J_0(2\pi f_m \cdot \Delta t)
$$

and

$$
S_{\mathbf{C}_{\ell}}(\lambda) = \begin{cases} \frac{1}{\pi f_m} \frac{1}{\sqrt{1 - (\lambda/f_m)^2}}, & |\lambda| \le f_m \\ 0, & \text{otherwise} \end{cases}
$$

where $\{$ ⎧⎪⎪⎪⎪⎪⎪⎪⎪⎪⎪⎪⎪⎪ $₁$ </sub> $f_m = (v/c)f_c$ is the maximum Doppler shift v is the vehicle speed (m/s) c is the light speed $(3 \times 10^8 \text{ m/s})$ f_c is the carrier frequency $J_0(\cdot)$ is the zero-order Bessel function of the first kind.

Jakes' model: Example 13.1-3

• Difference in path length

$$
\Delta L = \sqrt{(L\sin(\theta))^2 + (L\cos(\theta) + v \cdot \Delta t)^2} - L
$$

= $\sqrt{L^2 + v^2(\Delta t)^2 + 2L \cdot v \cdot \Delta t \cdot \cos(\theta)} - L$

• Phase change
$$
\Delta \phi = 2\pi \frac{\Delta L}{(c/f_c)}
$$
 $\left(= 2\pi \frac{\Delta L}{\text{ wavelength}} \right)$
• Estimated Doppler shift

$$
\lambda_m = \lim_{\Delta t \to 0} \frac{1}{2\pi} \frac{\Delta \phi}{\Delta t}
$$

=
$$
\frac{1}{c/f_c} \lim_{\Delta t \to 0} \frac{\sqrt{L^2 + v^2 (\Delta t)^2 + 2L \cdot v \cdot \Delta t \cdot \cos(\theta)} - L}{\Delta t}
$$

=
$$
\frac{v f_c}{c} \cos(\theta) = f_m \cos(\theta)
$$

Example. $v = 108$ km/hour, $f_c = 5$ GHz and $c = 1.08 \times 10^9$ km/hour.

$$
\implies \lambda_m = 500 \cos(\theta) \text{ Hz}.
$$

Notably, $\frac{500 \text{ Hz}}{5 \text{ GHz}} = 0.1 \text{ ppm}.$

Here, a rough derivation is provided for Jakes' model.

Just to give you a rough idea of how this model is obtained.

Suppose $\tau = \tau(t)$ is the delay of some path.

$$
\tau'(t) = \lim_{\Delta t \to 0} \frac{\tau(t + \Delta t) - \tau(t)}{\Delta t}
$$
\n
$$
= \lim_{\Delta t \to 0} \frac{\frac{t + \Delta t}{\Delta t} - \frac{t}{c}}{\frac{t}{\Delta t}}
$$
\n
$$
= \lim_{\Delta t \to 0} \frac{\Delta t}{c \Delta t}
$$
\n
$$
= \frac{v}{c} \cos(\theta)
$$
\n
$$
\Rightarrow \tau(t) \approx \frac{v}{c} \cos(\theta) t + \tau_0 \qquad v \leftarrow \frac{\sqrt{v} \sqrt{v}}{\sqrt{v}} \qquad (Assume for simplicity \tau_0 = 0.)
$$

Transmitter

Assume that $c(\tau; t) \approx a \cdot \delta(\tau - \tau(t))$, a constant-attenuation single-path system. Then

$$
c_{\ell}(\tau; t) = c(\tau; t) e^{-i 2\pi f_c \tau}
$$

\n
$$
\approx a \cdot \delta(\tau - \tau(t)) e^{-i 2\pi f_c \cdot \tau(t)}
$$

\n
$$
= a \cdot \delta(\tau - (v/c) \cos(\theta) t) e^{-i 2\pi f_c(\frac{v}{c} \cos(\theta) t)}
$$

\n
$$
= a \cdot \delta(\tau - (f_m/f_c) \cos(\theta) t) e^{-i 2\pi f_m \cos(\theta) t}
$$

and

$$
R_{c_{\ell}}(\tau; t + \Delta t, t)
$$
\n
$$
= \int_{-\infty}^{\infty} \mathbb{E} \left[c_{\ell}(\bar{\tau}; t + \Delta t) c_{\ell}^{*}(\tau; t) \right] d\bar{\tau}
$$
\n
$$
= \int_{-\infty}^{\infty} \mathbb{E} \left[a \cdot \delta(\bar{\tau} - (f_m/f_c) \cos(\theta)(t + \Delta t)) e^{-i 2\pi f_m \cos(\theta)(t + \Delta t)} \right]
$$
\n
$$
\cdot a \cdot \delta(\tau - (f_m/f_c) \cos(\theta)t) e^{i 2\pi f_m \cos(\theta)t} d\bar{\tau}
$$
\n
$$
= a^2 \cdot \mathbb{E} \left[e^{-i 2\pi f_m \cos(\theta) \cdot \Delta t} \right] \delta(\tau - (f_m/f_c) \cos(\theta)t)
$$

$$
R_{\mathbf{C}_{\ell}}(\Delta f = 0; t + \Delta t, t) \left(= \int_{-\infty}^{\infty} R_{c_{\ell}}(\tau; t + \Delta t, t) e^{i 2\pi (\Delta f)\tau} d\tau \right)_{\Delta f = 0}
$$

\n
$$
= \int_{-\infty}^{\infty} R_{c_{\ell}}(\tau; t + \Delta t, t) d\tau
$$

\n
$$
= \int_{-\infty}^{\infty} a^2 \cdot \mathbb{E} \left[e^{-i 2\pi f_m \cos(\theta) \cdot \Delta t} \right] \delta(\tau - (f_m/f_c) \cos(\theta) t) d\tau
$$

\n
$$
= a^2 \cdot \mathbb{E} \left[e^{-i 2\pi f_m \cos(\theta) \cdot \Delta t} \right]
$$

\n
$$
= J_0(2\pi f_m \cdot \Delta t) \left(= R_{\mathbf{C}_{\ell}}(\Delta f = 0; \Delta t) \right)
$$

where the last step is valid if θ uniformly distributed over $[-\pi, \pi)$, and $a = 1$.

 θ can be treated as uniformly distributed over $[-\pi, \pi)$ and independent of attenuation α and delay path τ .

Channel model from IEEE 802.11 Handbook

- A consistent channel model is required to allow comparison among different WLAN systems.¹
- The IEEE 802.11 Working Group adopted the following channel model as the baseline for predicting multipath for modulations used in IEEE 802.11a and IEEE 802.11b, which is ideal for software simulations.
	- The phase is uniformly distributed.
	- The magnitude is Rayleigh distributed with average power decaying exponentially.

 $1B.$ O'Hara and A. Petrick, IEEE 802.11 Handbook: A Designer's Companion, pp. 164–166, IEEE Press,1999

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Time Invariant:
$$
c_{\ell}(\tau; t) = c_{\ell}(\tau) = \sum_{i=0}^{i_{\text{max}}-1} \alpha_i e^{-i \phi_i} \delta(\tau - i \tau_s)
$$

$$
R_{c_{\ell}}(\tau) = \int_{-\infty}^{\infty} \mathbb{E} \left[c_{\ell}(\overline{\tau}; t) c_{\ell}^{*}(\tau; t) \right] d\overline{\tau} = \int_{-\infty}^{\infty} \mathbb{E} \left[c_{\ell}(\overline{\tau}) c_{\ell}^{*}(\tau) \right] d\overline{\tau}
$$

\n
$$
= \sum_{i=0}^{i_{\text{max}}-1} \int_{-\infty}^{\infty} \mathbb{E} \left[\alpha_{i}^{2} \right] \delta(\tau - i \tau_{s}) \delta(\overline{\tau} - \tau) d\overline{\tau}
$$

\n
$$
= \sum_{i=0}^{i_{\text{max}}-1} \mathbb{E} \left[\alpha_{i}^{2} \right] \delta(\tau - i \tau_{s})
$$

\n
$$
= \sum_{i=0}^{i_{\text{max}}-1} \sigma_{0}^{2} e^{-i \tau_{s} / \tau_{\text{rms}}} \delta(\tau - i \tau_{s})
$$

By this example, I want to introduce the rms delay spread. By definition, the "effective" rms delay is

$$
T_{\text{rms}}^{2} = \frac{\int_{-\infty}^{\infty} \tau^{2} R_{c_{\ell}}(\tau) d\tau}{\int_{-\infty}^{\infty} R_{c_{\ell}}(\tau) d\tau} - \left(\frac{\int_{-\infty}^{\infty} \tau R_{c_{\ell}}(\tau) d\tau}{\int_{-\infty}^{\infty} R_{c_{\ell}}(\tau) d\tau} \right)^{2}
$$

$$
= \frac{\sum_{i=0}^{i_{\text{max}}-1} (iT_{s})^{2} \sigma_{0}^{2} e^{-iT_{s}/\tau_{\text{rms}}}}{\sum_{i=0}^{i_{\text{max}}-1} \sigma_{0}^{2} e^{-iT_{s}/\tau_{\text{rms}}}} - \left(\frac{\sum_{i=0}^{i_{\text{max}}-1} (iT_{s}) \sigma_{0}^{2} e^{-iT_{s}/\tau_{\text{rms}}}}{\sum_{i=0}^{i_{\text{max}}-1} \sigma_{0}^{2} e^{-iT_{s}/\tau_{\text{rms}}}} \right)^{2}
$$

We wish to choose i_{max} such that $T_{\text{rms}} \approx \tau_{\text{rms}}$. Let $\tilde{\tau}_{\text{rms}} = \frac{\tau_{\text{rms}}}{T_s}$ and $\tilde{T}_{\text{rms}} = \frac{T_{\text{rms}}}{T_s}$. These unit-less terms $\tilde{\tau}_{\text{rms}}$ and \tilde{T}_{rms} are usually ≥ 1 .

$$
\tilde{\tau}_{\text{rms}}^2 = \frac{\sum_{i=0}^{n-1} i^2 p^i}{\sum_{i=0}^{n-1} p^i} - \left(\frac{\sum_{i=0}^{n-1} i p^i}{\sum_{i=0}^{n-1} p^i}\right)^2 \quad \text{with } p = e^{-1/\tilde{\tau}_{\text{rms}}} = e^{-x} \text{ and } n = i_{\text{max}}
$$
\n
$$
= \frac{p}{(1-p)^2} - \frac{n^2 p^n}{(1-p^n)^2} \quad \text{(Note } p^n = e^{-nx}.)
$$
\n
$$
= \left(\frac{1}{x^2} - \frac{1}{12} + \frac{x^2}{240} + \cdots\right) - \left(\frac{(nx)^2 e^{-nx}}{(1 - e^{-nx})^2}\right) \frac{1}{x^2} \approx \tilde{\tau}_{\text{rms}}^2 = \frac{1}{x^2}
$$
\n
$$
\text{where Taylor expansion yields } \frac{x^2 p}{(1-p)^2} = \frac{x^2 e^{-x}}{(1-e^{-x})^2} = 1 - \frac{x^2}{12} + \frac{x^4}{240} + O(x^8).
$$

$$
\frac{(nx)^2 e^{-nx}}{(1 - e^{-nx})^2} \le 0.01 \Rightarrow nx \ge 9 \Rightarrow i_{\text{max}} = n \ge \frac{9}{x} = 9\tilde{\tau}_{\text{rms}} = 9\frac{\tau_{\text{rms}}}{T_s}
$$

The Handbook suggests $i_{\text{max}} = 10$.

Typical multipath delay spread for indoor environment (Table 8-1 in IEEE 802.11 Handbook) with $T_s = 1/(20 \times 10^6) = 50$ nsec.

[13.1-2 Statistical models for fading](#page-46-0) [channels](#page-46-0)

In addition to zero-mean Gaussian (Rayleigh), non-zero-mean Gaussian (Rice) and Nakagami-m distributions, there are other models for $c_{\ell}(\tau; t)$ proposed in the literature.

Example.

Channels with a direct path and a single multipath component, such as airplane-to-ground communications

$$
c_{\ell}(\tau;t) = \alpha \delta(\tau) + \beta(t) \delta(\tau - \tau_0(t))
$$

where α controls the power in the direct path and is named *specular component*, and $\beta(t)$ is modeled as zero-mean Gaussian.

Example.

Microwave LOS radio channels used for long-distance voice and video transmission by telephone companies in the 6 GHz band (Rummler 1979)

$$
c_{\ell}(\tau) = \alpha \left[\delta(\tau) - \beta e^{i 2\pi f_0 \tau} \delta(\tau - \tau_0) \right]
$$

where

 α overall attenuation parameter

- β shape parameter due to multipath components
- τ_0 time delay

⎧⎪⎪⎪⎪⎪⎪⎪ ⎨ ⎪⎪⎪⎪⎪⎪⎪⎩ f_0 frequency of the fade minimum, i.e.,

$$
f_0 = \arg\min_{f \in \mathbb{R}} |\mathbf{C}_{\ell}(f)| = \arg\min_{f \in \mathbb{R}} \left|1 - \beta e^{-i 2\pi (f - f_0)\tau_0}\right|
$$

and $\operatorname{R}_{\ell}(f_0) = \operatorname{S}_{\ell}(f_0)\mathbf{C}_{\ell}(f_0) = \operatorname{S}_{\ell}(f_0)\alpha(1 - \beta).$

Rummler found that

- $\bullet \ \alpha \perp \beta$ (Independent)
- **2** $f(\beta) \approx (1-\beta)^{2.3}$ (pdf)
- \bullet -log(α) Gaussian distributed (i.e., α lognormal distributed)
- 4 $\tau_0 \approx 6.3$ ns

FIGURE 13.1-9 Magnitude frequency response of LOS channel model.

Deep fading phenomenon: At $f = f_0$, the so-called **deep fading** occurs.

[13.2 The effect of signal](#page-50-0) [characteristics on the choice of a](#page-50-0) [channel model](#page-50-0)

Usually, we prefer slowly fading and frequency non-selectivity.

So we wish to choose symbol time T and transmission bandwidth B such that

$$
T < (\Delta t)_c \quad \text{and} \quad B < (\Delta f)_c
$$

Hence, using $BT = 1$, we wish

$$
\frac{T}{(\Delta t)_c}\frac{B}{(\Delta f)_c}=B_d T_m<1.
$$

The term B_dT_m is an essential channel parameter and is called spread factor.

Underspread versus overspread

Underspread≡ $B_dT_m < 1$ Overspread≡ $B_dT_m > 1$

MULTIPATH SPREAD, DOPPLER SPREAD, AND SPREAD FACTOR FOR SEVERAL TIME-VARIANT MULTIPATH CHANNELS

[13.3 Frequency-nonslective, slowly](#page-53-0) [fading channel](#page-53-0)

For a frequency-nonslective, slowly fading channel, i.e.,

$$
T_m \ll \frac{1}{B} = T \ll (\Delta t)_c,
$$

the signal spectrum $s_{\ell}(f)$ is almost unchanged by $C_{\ell}(f; t)$; hence,

 $\mathbf{C}_\ell(f; t) \approx \mathbf{C}_\ell(0; t)$ within the signal bandwidth

and it is almost time-invariant; hence,

 $\mathbf{C}_\ell(f; t) \approx \mathbf{C}_\ell(0)$ within the signal bandwidth

This gives

$$
r_{\ell}(t) = c_{\ell}(\tau; t) \star s_{\ell}(t) + z(t)
$$

=
$$
\int_{-\infty}^{\infty} \mathbf{C}_{\ell}(f; t) s_{\ell}(f) e^{-i 2\pi t t} df + z(t)
$$

$$
\approx \int_{-\infty}^{\infty} \mathbf{C}_{\ell}(0) s_{\ell}(f) e^{-i 2\pi t t} df + z(t) = \mathbf{C}_{\ell}(0) s_{\ell}(t) + z(t)
$$

Assume that the phase of $\mathbf{C}_{\ell}(0) = \alpha e^{i\phi}$ can be perfectly
estimated and componented by the resolver. The channel estimated and compensated by the receiver. The channel model becomes:

$$
r_{\ell}(t)=\alpha s_{\ell}(t)+z(t).
$$

After demodulation (i.e., vectorization), we obtain

 $r_{\ell} = \alpha s_{\ell} + n_{\ell}.$

Question: What will the error probability be under random α ?

Case 1: Equal-prior BPSK

 $r = \pm \alpha \sqrt{\mathcal{E}} + n$ (passband vectorization with $E[n^2] = \frac{N_0}{2}$) \blacksquare $r_{\ell,real} = \pm \alpha \sqrt{2\mathcal{E}} + n_{\ell,real}$ (baseband vectorization with $E[n_{\ell,real}^2] = N_0$)

The optimal decision is $r \leq 0$, regardless of α (due to equal prior probability).

Thus,

$$
Pr\{\text{error}|\alpha\} = Q\left(\sqrt{2\alpha^2 \frac{\mathcal{E}}{N_0}}\right)
$$

Given that α is Rayleigh distributed (cf. Slide 4-167), we have

$$
P_{e,BPSK} = \int_0^\infty \Pr\{\text{error}|\alpha\} \underbrace{\frac{\alpha}{\sigma^2} e^{-\frac{\alpha^2}{2\sigma^2}} d\alpha}_{\text{Rayleigh}}
$$

Where
$$
\mathbb{E}[\alpha^2] = 2\sigma^2
$$
.

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$$
P_{e,BPSK} = \int_0^\infty Q(\beta \frac{\alpha}{\sigma}) \frac{\alpha}{\sigma^2} e^{-\frac{\alpha^2}{2\sigma^2}} d\alpha, \text{ where } \beta^2 = 2\sigma^2 \frac{\mathcal{E}}{N_0}
$$

\n
$$
= \int_0^\infty Q(\beta x) x e^{-\frac{x^2}{2}} dx, \text{ where } x = \frac{\alpha}{\sigma}
$$

\n
$$
= Q(\beta x) (-e^{-\frac{x^2}{2}}) \Big|_0^\infty - \int_0^\infty (-\frac{\beta}{\sqrt{2\pi}} e^{-\frac{\beta^2 x^2}{2}}) (-e^{-\frac{x^2}{2}}) dx
$$

\n
$$
= \frac{1}{2} - \sqrt{\frac{\beta^2}{1+\beta^2}} \int_0^\infty \frac{1}{\sqrt{2\pi(\frac{1}{1+\beta^2})}} e^{-\frac{x^2}{2(\frac{1}{1+\beta^2})}} dx
$$

\n
$$
= \frac{1}{2} - \frac{1}{2} \sqrt{\frac{\beta^2}{1+\beta^2}}
$$

\n
$$
= \frac{1}{2} - \frac{1}{2} \sqrt{\frac{\bar{\gamma}_b}{1+\bar{\gamma}_b}}, \text{ where } \bar{\gamma}_b = \mathbb{E}[\alpha^2] \frac{\mathcal{E}}{N_0} = (2\sigma^2)(\frac{\beta^2}{2\sigma^2}) = \beta^2
$$

\n
$$
= \frac{1}{2(1+\bar{\gamma}_b+\sqrt{\bar{\gamma}_b^2+\bar{\gamma}_b})} \approx \frac{1}{4\bar{\gamma}_b} \text{ when } \bar{\gamma}_b \text{ large}
$$

Case 2: Equal-prior BFSK

Similarly, for BFSK,

$$
\boldsymbol{r} = \left\{ \begin{bmatrix} \alpha \sqrt{\mathcal{E}} \\ 0 \end{bmatrix} \text{ or } \begin{bmatrix} 0 \\ \alpha \sqrt{\mathcal{E}} \end{bmatrix} \right\} + \boldsymbol{n}
$$

Under equal prior, the optimal decision is $r_1 \le r_2$, regardless of α.

$$
P_{e,BFSK} = \int_0^\infty \Pr\{\text{error}|\alpha\} f(\alpha) d\alpha
$$

\n
$$
= \int_0^\infty Q(\beta \frac{\alpha}{\sigma}) \frac{\alpha}{\sigma^2} e^{-\frac{\alpha^2}{2\sigma^2}} d\alpha, \text{ where now } \beta^2 = \sigma^2 \frac{\varepsilon}{N_0}
$$

\n
$$
= \frac{1}{2} - \frac{1}{2} \sqrt{\frac{\beta^2}{1+\beta^2}}
$$

\n
$$
= \frac{1}{2} - \frac{1}{2} \sqrt{\frac{\bar{\gamma}_b}{2+\bar{\gamma}_b}}, \text{ where } \bar{\gamma}_b = \mathbb{E}[\alpha^2] \frac{\varepsilon}{N_0} = 2\sigma^2 \frac{\beta^2}{\sigma^2} = 2\beta^2
$$

\n
$$
\left(= \frac{1}{2 + \bar{\gamma}_b + \sqrt{\bar{\gamma}_b^2 + 2\bar{\gamma}_b}} \approx \frac{1}{2\bar{\gamma}_b} \text{ when } \bar{\gamma}_b \text{ large} \right)
$$

Case 3: BDPSK

$$
P_{e,BDPSK} = \int_0^\infty \Pr\{\text{error}|\alpha\} f(\alpha) d\alpha
$$

\n
$$
= \int_0^\infty \left(\frac{1}{2} e^{-\beta^2 \frac{\alpha^2}{2\sigma^2}}\right) \left(\frac{\alpha}{\sigma^2} e^{-\frac{\alpha^2}{2\sigma^2}}\right) d\alpha, \text{ where } \beta^2 = 2\sigma^2 \frac{\mathcal{E}}{N_0}
$$

\n
$$
= \frac{1}{2(1+\beta^2)} \int_0^\infty (1+\beta^2) e^{-\left(\frac{1+\beta^2}{2}\right)x^2} dx
$$

\n
$$
= -\frac{1}{2(1+\beta^2)} e^{-\left(\frac{1+\beta^2}{2}\right)x^2} \Big|_0^\infty
$$

\n
$$
= \frac{1}{2(1+\bar{\gamma}_b)}, \text{ where } \bar{\gamma}_b = \mathbb{E}[\alpha^2] \frac{\mathcal{E}}{N_0} = 2\sigma^2 \frac{\beta^2}{2\sigma^2} = \beta^2
$$

\n
$$
\approx \frac{1}{2\bar{\gamma}_b} \text{ when } \bar{\gamma}_b \text{ large}
$$

$$
P_{e,noncoherent BFSK} = \int_0^\infty Pr\{\text{error}|\alpha\} f(\alpha) d\alpha
$$

\n
$$
= \int_0^\infty \left(\frac{1}{2} e^{-\beta^2 \frac{\alpha^2}{2\sigma^2}}\right) \left(\frac{\alpha}{\sigma^2} e^{-\frac{\alpha^2}{2\sigma^2}}\right) d\alpha, \text{ where } \beta^2 = \sigma^2 \frac{\varepsilon}{N_0}
$$

\n
$$
= \frac{1}{2(1+\beta^2)}
$$

\n
$$
= \frac{1}{2+\bar{\gamma}_b}, \text{ where } \bar{\gamma}_b = \mathbb{E}[\alpha^2] \frac{\varepsilon}{N_0} = 2\sigma^2 \frac{\beta^2}{\sigma^2} = 2\beta^2
$$

\n
$$
\approx \frac{1}{\bar{\gamma}_b} \text{ when } \bar{\gamma}_b \text{ large}
$$

● BPSK is 3dB better than BDPSK/BFSK; 6dB better than noncoherent BFSK.

 \bullet P_e decreases inversely proportional with SNR under fading.

 \bullet P_e decreases exponentially with SNR when **no** fading.

- To achieve $P_e = 10^{-4}$,
the system must provide the system must provide an SNR higher than 35dB, which is not practically possible!
- So an alternative solution should be used to compensate the fading such as the diversity technique.

If α \equiv Nakagami-*m* fading,

Turin et al. (1972) and Suzuki (1977) have shown that the Nakagami-m distribution is the best-fit for urban radio multipath channels.

$$
\implies f(\alpha) = \frac{2}{\Gamma(m)} \left(\frac{m}{\Omega}\right)^m \alpha^{2m-1} e^{-m\alpha^2/\Omega}, \text{ where } \Omega = \mathbb{E}[\alpha^2].
$$
\n• *m* < 1: Worse than Rayleigh fading in performance\n• *m* = 1: Rayleigh fading\n• *m* > 1: Better than Rayleigh fading in performance\nNotably, *m* =
$$
\frac{\mathbb{E}^2[\alpha^2]}{\text{Var}[\alpha^2]} = \frac{\Omega^2}{\mathbb{E}[(\alpha^2 - \Omega)^2]}
$$
 is called the fading figure.

Prob density function of Nakagami- m with $\Omega = 1$

BPSK performance under Nakagami-m fading

$$
P_{e,BPSK} = \int_{0}^{\infty} Q\left(\sqrt{2\alpha^{2} \mathcal{E}/N_{0}}\right)
$$
\n
$$
\frac{2}{\Gamma(m)} \left(\frac{m}{\Omega}\right)^{m} \alpha^{2m-1} e^{-m\alpha^{2}/\Omega} d\alpha
$$
\n
$$
f(\alpha)
$$
\n
$$
\frac{10^{-1}}{\frac{20}{\pi}} \left[\frac{10^{-3}}{\frac{20}{\pi}}\right]_{0}^{\frac{m}{2}} \left[\frac{10^{-3}}{\frac{20}{\pi}}\right]_{0}^{\frac{m}{2}}
$$
\n
$$
\frac{10^{-7}}{\frac{20}{\pi}} \left[\frac{10^{-5}}{\frac{20}{\pi}}\right]_{0}^{\frac{m}{2}} \left[\frac{10^{-5}}{\frac{20}{\pi}}\right]_{0}^{\frac{m}{2}} \left[\frac{10^{-5}}{\frac{20}{\pi}}\right]_{0}^{\frac{m}{2}} \left[\frac{10^{-5}}{\frac{20}{\pi}}\right]_{0}^{\frac{m}{2}} \left[\frac{10^{-5}}{\frac{20}{\pi}}\right]_{0}^{\frac{m}{2}} \left[\frac{10^{-5}}{\frac{20}{\pi}}\right]_{0}^{\frac{m}{2}} \left[\frac{10^{-5}}{\pi}\right]_{0}^{\frac{m}{2}} \left[\frac{10^{-5}}{\pi}\right]_{0}^{\frac{
$$

In some channel, the system performance may degrade even worse, such as Rummler's model in Slide 13-49, where deep fading occurs at some frequency.

The lowest is equal to $\alpha(1-\beta)$, which is itself a random variable.

[13.4 Diversity techniques for fading](#page-68-0) [multipath channels](#page-68-0)

Solutions to compensate deep fading

- Frequency diversity
	- Separation of carriers $\geq (\Delta f)_c = 1/T_m$ to obtain uncorrelation in signal replicas.
- Time diversity
	- Separation of time slots $\geq (\Delta t)_c = 1/B_d$ to obtain uncorrelation in signal replicas.
- Space diversity (Multiple receiver antennas)
	- Spaced sufficiently far apart to ensure received signals faded independently (usually, > 10 wavelengths)
- RAKE correlator or RAKE matched filter (Price and Green 1958)
	- It is named wideband approach, since it is usually applied to the situation where signal bandwidth is much greater than the coherent bandwidth $(\Delta f)_c$.

It is clear for the first three diversities, we will have L identical replicas at the Rx (which are uncorrelated).

The idea is that as long as not all of them are deep-faded, the demodulation is sufficiently good.

For the last one (i.e., RAKE), where $B \gg (\Delta f)_{c}$, which results in a frequency selective channel, we have

$$
L=\frac{B}{(\Delta f)_c}.
$$

Detail will be given in the following.

[13.4-1 Binary signals](#page-71-0)
Assumption

- **1** L identical and independent channels
- 2 Each channel is frequency-nonselective and slowly fading with Rayleigh-distributed envelope.
- ³ Zero-mean additive white Gaussian background noise
- **4** Assume the phase-shift can be perfectly compensated.
- **B** Assume the attenuation $\{\alpha_k\}_{k=1}^L$ can be perfectly estimated at Rx.

Hence,

$$
r_k = \alpha_k s + n_k \quad k = 1, 2, \dots, L
$$

How to combine these L outputs when making decision? Maximal ratio combiner (Brennan 1959)

$$
r = \sum_{k=1}^L \alpha_k r_k = \sum_{k=1}^L \alpha_k^2 s + \sum_{k=1}^L \alpha_k n_k
$$

Idea behind maximal ratio combiner

• Trust more on the strong signals and trust less on the weak signal.

Advantage of maximal ratio combiner

- Theoretically tractable; so we can predict how "good" the system can achieve without performing simulations.
- Maximum-ratio combining is an optimum linear combiner.

Find $\{w_k\}_{k=1}^L$ for linear combiner

$$
r = \sum_{k=1}^{L} w_k r_k = \sum_{k=1}^{L} w_k \alpha_k s + \sum_{k=1}^{L} w_k n_k
$$

such that the output SNR

$$
\left(\sum_{k=1}^{L} w_k \alpha_k\right)^2 \mathbb{E}[s^2] / \sum_{k=1}^{L} w_k^2 \mathbb{E}[n_k^2]
$$

is maximized.

Case 1: Equal-prior BPSK

$$
r = \pm \alpha \sqrt{\mathcal{E}} + n
$$
, where $\alpha = \sqrt{\sum_{k=1}^{L} \alpha_k^2}$ and $n = \frac{1}{\alpha} \sum_{k=1}^{L} \alpha_k n_k$

The optimal decision is $r \leq 0$, regardless of α . Thus,

$$
Pr\{\text{error}|\alpha\} = Q\left(\sqrt{2\alpha^2 \frac{\mathcal{E}}{N_0}}\right)
$$

Given that $\{\alpha_k\}_{k=1}^L$ is i.i.d. Rayleigh distributed, α is Nakagami-L distributed; hence,

$$
P_{e,BPSK} = \int_0^\infty \Pr\{\text{error}|\alpha\} \frac{2}{(L-1)!} \left(\frac{L}{\Omega}\right)^L \alpha^{2L-1} e^{-L\alpha^2/\Omega} d\alpha
$$
\nwhere $\Omega = \mathbb{E}[\alpha^2] = L \cdot \mathbb{E}[\alpha_1^2] = 2L\sigma^2$.
\nNote $\{\alpha_k^2 = X_k^2 + Y_k^2\}_{k=1}^L$ is i.i.d. χ^2 -distributed with 2 degree of freedom
\n $\Rightarrow \alpha^2 = \sum_{k=1}^L \alpha_k^2 = X_1^2 + Y_1^2 + \dots + X_L^2 + Y_L^2$ is χ^2 -distributed with 2L degree
\nof freedom, where $\{X_k\}_{k=1}^L$ and $\{Y_k\}_{k=1}^L$ zero-mean i.i.d. Gaussian.

$$
P_{e,BPSK} = \int_0^\infty Q(\beta \alpha) \frac{2}{(L-1)!} \left(\frac{1}{2\sigma^2}\right)^L \alpha^{2L-1} e^{-\alpha^2/(2\sigma^2)} d\alpha,
$$

\nwhere $\beta = \sqrt{2\mathcal{E}/N_0}$
\n
$$
= \left(\frac{1-\mu}{2}\right)^L \cdot \sum_{k=0}^{L-1} {L-1+k \choose k} \left(\frac{1+\mu}{2}\right)^k
$$

\nwhere $\mu = \sqrt{\frac{\bar{\gamma}_c}{1+\bar{\gamma}_c}}$ and $\bar{\gamma}_c = \mathbb{E}[\alpha_k^2] \frac{\mathcal{E}}{N_0}$
\n
$$
\left(\approx \left(\frac{2L-1}{L}\right) \left(\frac{1}{4\bar{\gamma}_c}\right)^L \text{ when } \bar{\gamma}_c \text{ large}\right)
$$

\nwhere we have $\frac{1-\mu}{2} = \frac{1}{2(1+\bar{\gamma}_c+\sqrt{\bar{\gamma}_c^2+\bar{\gamma}_c})} \approx \frac{1}{4\bar{\gamma}_c}$ and $\frac{1+\mu}{2} \approx 1$.

For your reference: $L = 2$

$$
P_{e,BPSK} = \int_{0}^{\infty} Q(\beta \frac{\alpha}{\sigma}) 2(\frac{1}{2\sigma^{2}})^{2} \alpha^{3} e^{-\frac{\alpha^{2}}{2\sigma^{2}}} d\alpha, \text{ where } \beta^{2} = 2\sigma^{2} \frac{\mathcal{E}}{N_{0}}
$$

\n
$$
= \int_{0}^{\infty} Q(\beta x) \frac{1}{2} x^{3} e^{-\frac{x^{2}}{2}} dx
$$

\n
$$
= Q(\beta x) \left(-\frac{(x^{2}+2)}{2} e^{-\frac{x^{2}}{2}} \right) \Big|_{0}^{\infty} - \int_{0}^{\infty} \left(-\frac{\beta}{\sqrt{2\pi}} e^{-\frac{\beta^{2}x^{2}}{2}} \right) \left(-\frac{(x^{2}+2)}{2} e^{-\frac{x^{2}}{2}} \right) dx
$$

\n
$$
= \frac{1}{2} - \int_{0}^{\infty} \frac{\beta}{2\sqrt{2\pi}} (x^{2}+2) e^{-\frac{(1+\beta^{2})x^{2}}{2}} dx
$$

\n
$$
= \frac{1}{2} - \frac{1}{4} \sqrt{\frac{\beta^{2}}{1+\beta^{2}}} \int_{-\infty}^{\infty} (x^{2}+2) \frac{1}{\sqrt{2\pi} \frac{1}{(1+\beta^{2})}} e^{-\frac{x^{2}}{2(1+\beta^{2})}} dx
$$

\n
$$
= \frac{1}{2} - \frac{1}{4} \sqrt{\frac{\beta^{2}}{1+\beta^{2}}} \left[\frac{1}{(1+\beta^{2})} + 2 \right]
$$

\n
$$
= \frac{1}{2} - \frac{1}{4} \mu ((1 - \mu^{2}) + 2), \text{ where } \bar{\gamma}_{c} = \mathbb{E}[\alpha_{k}^{2}] \frac{\mathcal{E}}{N_{0}} = \beta^{2} \text{ and } \mu = \sqrt{\frac{\bar{\gamma}_{c}}{1+\bar{\gamma}_{c}}}
$$

\n
$$
= \left(\frac{1 - \mu}{2} \right)^{2} \cdot \left(1 + 2 \left(\frac{1 + \mu}{2} \right)^{1} \right)
$$

Case 2: Equal-prior BFSK

Similarly, for BFSK,

$$
\mathbf{r} = \left\{ \begin{bmatrix} \alpha \sqrt{\mathcal{E}} \\ 0 \end{bmatrix} \text{ or } \begin{bmatrix} 0 \\ \alpha \sqrt{\mathcal{E}} \end{bmatrix} \right\} + \mathbf{n}
$$

The optimal decision is $r_1 \le r_2$, regardless of α .

$$
P_{e,BFSK} = \int_0^\infty Q(\beta \alpha) \frac{2}{(L-1)!} \left(\frac{1}{2\sigma^2}\right)^L \alpha^{2L-1} e^{-\alpha^2/(2\sigma^2)} d\alpha,
$$

where $\beta^2 = \sigma^2 \frac{\varepsilon}{N_0}$

$$
= \left(\frac{1-\mu}{2}\right)^{L} \cdot \sum_{k=0}^{L-1} {L-1+k \choose k} \left(\frac{1+\mu}{2}\right)^{k}
$$

where $\mu = \sqrt{\frac{\tilde{\gamma}_{c}}{2+\tilde{\gamma}_{c}}}$ and $\tilde{\gamma}_{c} = \mathbb{E}[\alpha_{k}^{2}] \frac{\mathcal{E}}{\mathcal{N}_{0}}$
 $\left(\approx \left(\frac{1}{2\tilde{\gamma}_{c}}\right)^{L} {2L-1 \choose L}$ when $\tilde{\gamma}_{c}$ large

Case 3: BDPSK

From Slide 4-175, the two consecutive lowpass equivalent signals are

$$
\mathbf{s}_{\ell}^{(k-1)} = \sqrt{2\mathcal{E}} e^{\imath \phi_0} \quad \text{and} \quad \mathbf{s}_{\ell}^{(k)} = \begin{cases} \sqrt{2\mathcal{E}} e^{\imath \phi_0}, & m = 1; \\ -\sqrt{2\mathcal{E}} e^{\imath \phi_0}, & m = 2 \end{cases}
$$

The L received signals given $\boldsymbol{s}_{\ell}^{(k-1)}$ and $\boldsymbol{s}_{\ell}^{(k)}$ are

$$
\vec{r}_{j,\ell} = \begin{bmatrix} \mathbf{r}_{j,\ell}^{(k-1)} \\ \mathbf{r}_{j,\ell}^{(k)} \end{bmatrix} = \alpha_j e^{\imath \phi_j} \begin{bmatrix} \mathbf{s}_{\ell}^{(k-1)} \\ \mathbf{s}_{\ell}^{(k)} \end{bmatrix} + \begin{bmatrix} \mathbf{n}_{j,\ell}^{(k-1)} \\ \mathbf{n}_{j,\ell}^{(k)} \end{bmatrix} = \alpha_j e^{\imath \phi_j} \vec{\mathbf{s}}_{\ell} + \vec{\mathbf{n}}_{j,\ell}
$$

for $j = 1, \ldots, L$.

Note that it is unnecessary to estimate α_j and ϕ_j for *j*th reception as required by Cases 1 & 2.

$$
\Rightarrow \vec{s}_{j,\ell}^{\dagger} \vec{r}_{j,\ell} = \begin{bmatrix} \sqrt{2\mathcal{E}} e^{-i\phi_0} & \pm \sqrt{2\mathcal{E}} e^{-i\phi_0} \end{bmatrix} \begin{bmatrix} \mathbf{r}_{j,\ell}^{(k-1)} \\ \mathbf{r}_{j,\ell}^{(k)} \end{bmatrix}
$$

$$
= \begin{cases} \sqrt{2\mathcal{E}} e^{-i\phi_0} (\mathbf{r}_{j,\ell}^{(k-1)} + \mathbf{r}_{j,\ell}^{(k)}), & m=1 \\ \sqrt{2\mathcal{E}} e^{-i\phi_0} (\mathbf{r}_{j,\ell}^{(k-1)} - \mathbf{r}_{j,\ell}^{(k)}), & m=2 \end{cases}
$$
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Instead of maximal ratio combining, we do square-law combining:

$$
\hat{m} = \arg \max_{1 \le m \le 2} \sum_{j=1}^{L} \left| \vec{s}_{j,\ell}^{\dagger} \vec{r}_{j,\ell} \right|^{2}
$$
\n
$$
= \arg \max \left\{ \underbrace{\sum_{j=1}^{L} \left| \vec{r}_{j,\ell}^{(k-1)} + \vec{r}_{j,\ell}^{(k)} \right|^{2}}_{m=1}, \underbrace{\sum_{j=1}^{L} \left| \vec{r}_{j,\ell}^{(k-1)} - \vec{r}_{j,\ell}^{(k)} \right|^{2}}_{m=2} \right\}
$$
\n
$$
= \arg \max \left\{ \underbrace{U_{\ell}}_{m=1}, \underbrace{-U_{\ell}}_{m=2} \right\}
$$

where
$$
U_{\ell} = \sum_{j=1}^{L} \textbf{Re} \left\{ \left(\mathbf{r}_{j,\ell}^{(k-1)} \right)^{*} \mathbf{r}_{j,\ell}^{(k)} \right\}.
$$

The quadratic-form analysis (cf. Slide 4-176) gives

$$
P_{e,BDPSK} = \left(\frac{1-\mu}{2}\right)^{L} \cdot \sum_{k=0}^{L-1} {L-1+k \choose k} \left(\frac{1+\mu}{2}\right)^{k}
$$

with $\mu = \frac{\bar{\gamma}_c}{1+\bar{\gamma}_c}$

$$
\approx \left(\frac{1}{2\bar{\gamma}_c}\right)^{L} {2L-1 \choose L} \text{ when } \bar{\gamma}_c \text{ large.}
$$

Case 4: Noncoherent BFSK

Recall from Slide 4-165:

The noncoherent ML computes

$$
\hat{m} = \arg \max_{1 \leq m \leq 2} \left| \boldsymbol{r}_{\ell}^{\dagger} \boldsymbol{s}_{m,\ell} \right|
$$

$$
\begin{array}{cc}\n\mathbf{s}_{1,\ell} = & \left(\begin{array}{cc} \sqrt{2\mathcal{E}_s} & 0 \\ 0 & \sqrt{2\mathcal{E}_s} \end{array} \right)\n\\
\mathbf{s}_{2,\ell} = & \left(\begin{array}{cc} 0 & \sqrt{2\mathcal{E}_s} \end{array} \right)\n\end{array}
$$

Hence,

$$
\hat{m} = \arg \max_{1 \le m \le 2} \left| r_{m,\ell} \right| = \arg \max_{1 \le m \le 2} \left| r_{m,\ell} \right|^2
$$

Now we have L diversities/channels:

$$
\pmb{r}_{j,\ell} = \begin{bmatrix} r_{j,1,\ell} \\ r_{j,2,\ell} \end{bmatrix} = \alpha_j e^{\imath \phi_j} \pmb{s}_{m,\ell} + \pmb{n}_{j,\ell} \quad j = 1,2,\ldots,L
$$

Instead of maximal ratio combining, we again do square-law combining:

$$
\hat{m} = \arg \max_{1 \leq m \leq 2} \sum_{j=1}^{L} \left| \mathbf{r}_{j,m,\ell} \right|^2.
$$

$$
P_{e,noncoherent BFSK} = \left(\frac{1-\mu}{2}\right)^{L} \cdot \sum_{k=0}^{L-1} {L-1+k \choose k} \left(\frac{1+\mu}{2}\right)^{k}
$$

with $\mu = \frac{\bar{\gamma}_c}{2+\bar{\gamma}_c}$

$$
\approx \left(\frac{1}{\bar{\gamma}_c}\right)^{L} {2L-1 \choose L} \text{ when } \bar{\gamma}_c \text{ large.}
$$

Summary (what the theoretical results indicate?)

With Lth order diversity, the POE decreases inversely with Lth power of the SNR.

FIGURE 13.4-2 Performance of binary signals with diversity.

In Cases 1 & 2, comparing the prob density functions of α for 1-diversity (no diversity) **Nakagami fading** and L -diversity Rayleigh fading, we conclude:

L-diversity in Rayleigh fading $= 1$ -diversity in Nakagami-L

or further

 m L-diversity in Rayleigh fading $=$ L-diversity in Nakagami- m

[13.4-2 Multiphase signals](#page-85-0)

For M -ary phase signal over L Rayleigh fading channels, the symbol error rate P_e can be derived as (Appendix C)

$$
P_e = \frac{(-1)^{L-1}(1-\mu^2)^L}{\pi(L-1)!} \left(\frac{\partial^{L-1}}{\partial b^{L-1}} \left\{ \frac{1}{b-\mu^2} \left[\frac{\pi}{M}(M-1) - \frac{\mu \sin(\pi/M)}{\sqrt{b-\mu^2 \cos^2(\pi/M)}} \cot^{-1} \left(\frac{-\mu \cos(\pi/M)}{\sqrt{b-\mu^2 \cos^2(\pi/M)}} \right) \right] \right\}_{b=1}
$$

$$
\approx \left\{ \frac{\frac{M-1}{\log_2(M) \sin^2(\pi/M)} \frac{1}{2M\gamma_b}}{\frac{M-1}{\log_2(M) \sin^2(\pi/M)} \frac{1}{M\gamma_b}} \right. \text{ M-ary PSK & L=1}
$$

where

$$
\mu = \begin{cases} \sqrt{\frac{\bar{\gamma}_c}{1+\bar{\gamma}_c}} & \text{M-ary PSK} \\ \frac{\bar{\gamma}_c}{1+\bar{\gamma}_c} & \text{M-ary DPSK} \end{cases}
$$

and in this case, the system SNR $\bar{\gamma}_t = \bar{\gamma}_b \log_2(M) = L \bar{\gamma}_c$.

PSK is about 3dB better than DPSK for all M $(L = 1)$.

Recall Slide 4-180 under AWGN, BDPSK is 1 dB inferior than BPSK and QDPSK is 2.3 dB inferior than QPSK.

FIGURE 13.4-3 Probability of symbol error for PSK and DPSK for Rayleigh fading.

DPSK performance with diversity

- Bit error P_h is calculated based on Gray coding.
- Larger M , worse P_b except for equal P_b at $M = 2, 4$.

FIGURE 13.4-4 Probability of a bit error for DPSK with diversity for Rayleigh fading.

13.4-3 M[-ary orthogonal signals](#page-89-0)

Noncoherent detection

- Here, the derivation assumes that both passband and lowpass equivalent signals are orthogonal; hence, the frequency separation is $1/T$ rather than $1/(2T)$.
- Based on lowpass (baseband) orthogonality, L-diversity square-law combining gives

$$
P_e = \frac{1}{(L-1)!} \sum_{m=1}^{M-1} \frac{(-1)^{m+1} {M-1 \choose m}}{(1+m+m\overline{\gamma}_c)^L}
$$

$$
\sum_{k=0}^{m(L-1)} \beta_{k,m}(L-1+k)! \left(\frac{1+\overline{\gamma}_c}{1+m+m\overline{\gamma}_c}\right)^k
$$

where $\beta_{k,m}$ is the coefficient of U^k in $\left(\sum_{k=0}^{L-1} \frac{U^k}{k!}\right)$ $\overline{k!}$) m , i.e.,

$$
\left(\sum_{k=0}^{L-1} \frac{U^k}{k!}\right)^m = \sum_{k=0}^{m(L-1)} \beta_{k,m} U^k.
$$

 $M = 2$ case:

- Let $\bar{\gamma}_t = L\bar{\gamma}_c$ be the total system power. For fixed $\bar{\gamma}_t$, there is an L that minimizes P_{e} .
- This hints that

 $\bar{\gamma}_c$ = 3 \approx 4.77 dB

gives the best performance.

 $M = 4$ case:

- Let $\bar{\gamma}_t = L\bar{\gamma}_c$ be the total system power. For fixed $\bar{\gamma}_t$, there is an L that minimizes P_{e} .
- This hints that $\bar{\gamma}_c = 3 \approx 4.77$ dB

gives the best performance.

Discussions:

- Larger M , better performance but larger bandwidth.
- Larger L, better performance.
- \bullet An increase in \prime is more efficient in performance gain than an increase in M.

[13.5 Digital signaling over a](#page-94-0) [frequency-selective, slowly fading](#page-94-0) [channel](#page-94-0)

[13.5.1 A tapped-delay-line channel](#page-95-0) [model](#page-95-0)

Assumption (Time-invariant channel)

$$
c_{\ell}(\tau;t)=c_{\ell}(\tau)
$$

Assumption (Bandlimited signal)

 $s_{\ell}(t)$ is band-limited, i.e., $|s_{\ell}(f)| = 0$ for $|f| > W/2$

In such case, we shall add a lowpass filter at the Rx.

Equivalent channel with $C_{\ell}(f)$ random

Equivalent channel with $C_{\ell}(f)$ random and bandlimited, and $z_W(t)$ bandlimited white noise

$$
r_{\ell}(t) = \int_{-\infty}^{\infty} s_{\ell}(f) \mathbf{C}_{\ell}(f) e^{i 2\pi ft} df + z_W(t)
$$

For a bandlimited $C_{\ell}(f)$, sampling theorem gives:

$$
c_{\ell}(t) = \sum_{n=-\infty}^{\infty} c_{\ell} \left(\frac{n}{W}\right) \operatorname{sinc}\left(W\left(t - \frac{n}{W}\right)\right)
$$

$$
c_{\ell}(f) = \int_{-\infty}^{\infty} c_{\ell}(t) e^{-i2\pi ft} dt
$$

$$
= \begin{cases} \frac{1}{W} \sum_{n=-\infty}^{\infty} c_{\ell} \left(\frac{n}{W}\right) e^{-i2\pi fn/W}, & |f| \leq W/2\\ 0, & \text{otherwise} \end{cases}
$$

$$
r_{\ell}(t) = \int_{-\infty}^{\infty} s_{\ell}(f) \mathbf{C}_{\ell}(f) e^{i2\pi ft} df + z_{W}(t)
$$

\n
$$
= \frac{1}{W} \sum_{n=-\infty}^{\infty} c_{\ell} \left(\frac{n}{W}\right) \int_{-W/2}^{W/2} s_{\ell}(f) e^{i2\pi f(t-n/W)} df + z_{W}(t)
$$

\n
$$
= \frac{1}{W} \sum_{n=-\infty}^{\infty} c_{\ell} \left(\frac{n}{W}\right) s_{\ell} \left(t - \frac{n}{W}\right) + z_{W}(t)
$$

\n
$$
= \sum_{n=-\infty}^{\infty} c_{n} \cdot s_{\ell} \left(t - \frac{n}{W}\right) + z_{W}(t), \text{ where } c_{n} = \frac{1}{W} c_{\ell} \left(\frac{n}{W}\right)
$$

For a time-varying channel, we replace $c_{\ell}(\tau)$ and $\mathbf{C}_{\ell}(f)$ by $c_{\ell}(\tau; t)$ and $\mathbf{C}_{\ell}(f ; t)$ and obtain

$$
r_{\ell}(t) = \sum_{n=-\infty}^{\infty} c_n(t) \cdot s_{\ell}\left(t - \frac{n}{W}\right) + z_W(t)
$$

where $c_n(t) = \frac{1}{W} c_\ell \left(\frac{n}{W}; t \right)$.

Statistically, with probability one, $c_{\ell}(\tau) = 0$ for $\tau > T_m$ and $\tau < 0$.

So, $c_{\ell}(\tau)$ is assumed band-limited and is also statistically time-limited!

Hence, $c_n(t) = 0$ for $n < 0$ and $n > T_mW$ (since $\tau = n/W > T_m$).

$$
r_{\ell}(t) = \sum_{n=0}^{\lfloor T_m W \rfloor} c_n(t) \cdot s_{\ell} \left(t - \frac{n}{W} \right) + z_W(t)
$$

FIGURE 13.5-1 Trapped delay line model of frequency-selective channel. For convenience, the text re-indexes the system as

$$
r_{\ell}(t)=\sum_{k=1}^L c_k(t)\cdot s_{\ell}\left(t-\frac{k}{W}\right)+z_W(t).
$$

[13.5-2 The RAKE demodulator](#page-101-0)

Assumption (Gaussian and US (uncorrelated scattering))

 $\{c_k(t)\}_{k=1}^L$ complex i.i.d. Gaussian and can be perfectly estimated
by By by Rx.

So the Rx can regard the "transmitted signal" as one of

$$
\begin{cases}\nv_{1,\ell}(t) = \sum_{k=1}^{L} c_k(t) \cdot s_{1,\ell}(t - \frac{k}{W}) \\
\vdots \\
v_{M,\ell}(t) = \sum_{k=1}^{L} c_k(t) \cdot s_{M,\ell}(t - \frac{k}{W})\n\end{cases}
$$

So Slide 4-158 said:

Coherent MAP detection

$$
\hat{m} = \arg \max_{1 \le m \le M} \text{Re}\left[r_{\ell}^{\dagger} v_{m,\ell}\right] = \arg \max_{1 \le m \le M} \text{Re}\left[\int_{0}^{T} r_{\ell}(t) v_{m,\ell}^{*}(t) dt\right]
$$
\n
$$
= \arg \max_{1 \le m \le M} \text{Re}\left[\sum_{k=1}^{L} \int_{0}^{T} r_{\ell}(t) c_{k}^{*}(t) s_{m,\ell}^{*}\left(t - \frac{k}{W}\right) dt\right]
$$
\n
$$
= \frac{U_{m,\ell}}{U_{m,\ell}}
$$

Discussions on assumptions: We assume:

- \bullet s_e(t) is band-limited to W.
- \bullet $c_{\ell}(\tau)$ is causal and (statistically) time-limited to T_m and, at the same time, band-limited to W .

•
$$
W \gg (\Delta f)_c = \frac{1}{T_m}
$$
 (i.e., $L \approx WT_m \gg 1$)

• The definition of $U_{m,\ell}$ requires $T \gg T_m$ (See page 871 in textbook) such that the longest delayed version

$$
s_{\ell}(t-L/W)=s_{\ell}(t-WT_m/W)=s_{\ell}(t-T_m)
$$

is still well-confined within the integration range $[0, T)$. As a result, the signal bandwidth is much larger than $1/T$; RAKE is used in the demodulation of "spread-spectrum" signals!

$$
WT \gg L \approx WT_m \gg 1 \implies W \gg \frac{1}{7}
$$

The receiver collects the signal energy from all received paths, which is somewhat analogous to the garden rake that is used to gather leaves, hays, etc. Consequently, the name "RAKE receiver" has been coined for this receiver structure by Price and Green (1958) . (I use $s_{m,\ell}$, but the text uses $s_{\ell,m}$.)

An alternative realization of RAKE receiver

The previous structure requires M delay lines.

We can reduce the number of the delay lines to **one** by the following derivation.

Let
$$
u = t - \frac{k}{W}
$$
.
\n
$$
U_{m,\ell} = \text{Re}\left[\sum_{k=1}^{L} \int_{0}^{T} r_{\ell}(t) c_{k}^{*}(t) s_{m,\ell}^{*} \left(t - \frac{k}{W}\right) dt\right]
$$
\n
$$
= \text{Re}\left[\sum_{k=1}^{L} \int_{-k/W}^{T-k/W} r_{\ell}\left(u + \frac{k}{W}\right) c_{k}^{*} \left(u + \frac{k}{W}\right) s_{m,\ell}^{*}(u) du\right]
$$
\n
$$
\approx \text{Re}\left[\sum_{k=1}^{L} \int_{0}^{T} r_{\ell}\left(t + \frac{k}{W}\right) c_{k}^{*} \left(t + \frac{k}{W}\right) s_{m,\ell}^{*}(t) dt\right]
$$
\nwhere the last approximation follows from\n
$$
\left|\frac{k}{W}\right| \leq \left|\frac{L}{W}\right| \approx \left|\frac{T_{m}W}{W}\right| = T_{m} \ll T \text{ (See Slide 13-104)}.
$$
\n
$$
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$$
\n
$$
\frac{106}{118}
$$

FIGURE 13.5-3 Optimum demodulator for wideband binary signals (delayed received signal configuration).

 $c_k^*\left(t+\frac{k}{W}\right)=\frac{1}{W}c_\ell^*\left(\frac{k}{W};t+\frac{k}{W}\right)$ is abbreviated as $c_k^*(t)$ in the above figure.

Performance of RAKE receiver

Suppose $c_k(t) = c_k$ and the signal corresponding to $m = 1$ is transmitted. Then, letting $\tilde{U}_{m,\ell} = \frac{1}{\sqrt{2\varepsilon_s}} U_{m,\ell}$ and $2\mathcal{E}_s$ $\tilde{s}^*_{m,\ell}$ ($t - \frac{k}{W}$ $\frac{k}{W}$) = $\frac{1}{\sqrt{2}}$ $\frac{1}{2\varepsilon_s} s_{m,\ell}^* \left(t - \frac{k}{W} \right)$ $\frac{\kappa}{W}$) (normalization), we have $\tilde{U}_{m,\ell}$ = Re $\Big[\sum_{k=1}^{L}$ ∑ $\sum_{k=1}$ ∫ T $\int_0^T r_\ell(t) c_k^* \tilde{s}_{m,\ell}^* \bigg(t - \frac{k}{W}\bigg)$ $\frac{1}{W}$ ^{dt} $=$ Re ▎ ⎣ L ∑ $\sum_{k=1}$ T 0 (L $\sum_{n=1}^{L} c_n s_{1,\ell} \left(t - \frac{n}{W} \right)$ $\frac{1}{W}$ + z_W(t) $\Big)$ $m = 1$ $c_k^* \tilde{s}_{m,\ell}^* \left(t - \frac{k}{\mathsf{M}} \right)$ $\frac{1}{W}$ ^{dt} \Box $=$ Re L ∑ $k=1$ L $\sum_{n=1}$ $c_n c_k^*$ T $\int_0^T s_{1,\ell}\bigg(t-\frac{n}{W}\bigg)$ $\left(\frac{n}{W}\right)$ s̃*, $\ell\left(t-\frac{k}{W}\right)$ $\frac{1}{W}$ ^{dt} $+$ Re $\Big|$ L ∑ $k=1$ c_k^* \int T $\int_0^T z_W(t) \tilde{s}^*_{m,\ell} \bigg(t - \frac{k}{W}\bigg)$ $\frac{1}{W}$ ^{dt}
Assumption (Add-and-delay property)

The transmitted signal is orthogonal to the shifted counterparts of all signals, including itself.

•
$$
\begin{cases} z_k = \int_0^T z_W(t) \tilde{s}_{m,\ell}^* \left(t - \frac{k}{W} \right) dt \Big|_{k=1}^L \text{ complex Gaussian with} \\ E[|z_k|^2] = 2N_0 \text{ because } \{ \tilde{s}_{m,\ell}^* \left(t - \frac{k}{W} \right) \}_{k=1}^L \text{ orthonormal.} \end{cases}
$$

Hence, with $\alpha_k = |c_k|$,

$$
\tilde{U}_{m,\ell} = \text{Re}\left[\sum_{k=1}^{L} |c_k|^2 \int_0^T s_{1,\ell} \left(t - \frac{k}{W}\right) \tilde{s}_{m,\ell}^* \left(t - \frac{k}{W}\right) dt\right] + \text{Re}\left[\sum_{k=1}^{L} c_k^* z_k\right]
$$
\n
$$
= \sum_{k=1}^{L} \alpha_k^2 \text{Re}\left[\left\langle s_{1,\ell} \left(t - \frac{k}{W}\right), \tilde{s}_{m,\ell} \left(t - \frac{k}{W}\right) \right\rangle\right] + \sum_{k=1}^{L} \alpha_k n_{k,\ell},
$$

where $\{n_{k,\ell} = \textbf{Re}[e^{-i\angle c_k}z_k]\}_{k=1}^L$ i.i.d. Gaussian with $E[n_{k,\ell}^2] = N_0$.

Under $T \gg T_m$, \int_0^T $\int_0^t S_{1,\ell}(t - \frac{k}{W})$ $\frac{k}{W}$) $\tilde{s}_{m,\ell}^*$ (t – $\frac{k}{W}$ $\frac{k}{W}$)dt is almost functionally independent of k ; so,

$$
\left\langle s_{1,\ell}\left(t-\frac{k}{W}\right),\tilde{s}_{m,\ell}\left(t-\frac{k}{W}\right)\right\rangle \approx \left\langle s_{1,\ell}\left(t\right),\tilde{s}_{m,\ell}\left(t\right)\right\rangle.
$$

Therefore, the performance of RAKE is the same as the L-diversity maximal ratio combiner if $\{\alpha_k\}_{k=1}^L$ i.i.d.
However, $\{\alpha_k = |\alpha| \}$ may not be identified However, $\{\alpha_k = |c_k|\}_{k=1}^L$ may not be identically distributed.

In such case, we can still obtain the pdf of $\gamma_b = \sum_{k=1}^L \gamma_k = \sum_{k=1}^L \alpha_k^2 \mathcal{E}_s / N_0 = \alpha^2 \mathcal{E}_s / N_0$ from

$$
\begin{cases}\n\text{characteristic function of } \gamma_k \equiv \Psi_k(\iota \nu) = \frac{1}{1 - \iota \nu \bar{\gamma}_k} \\
\text{characteristic function of } \gamma_b = \sum_{k=1}^L \gamma_k \equiv \prod_{k=1}^L \Psi_k(\iota \nu) = \prod_{k=1}^L \frac{1}{1 - \iota \nu \bar{\gamma}_k}\n\end{cases}
$$

The pdf of γ_b is then given by the Fourier transform of characteristic function:

$$
f(\gamma_b) = \sum_{k=1}^L \frac{\pi_k}{\bar{\gamma}_k} e^{-\gamma_b/\bar{\gamma}_k}
$$

where with
$$
\bar{\gamma}_k = \mathbb{E}[\gamma_k]
$$
, $\pi_k = \prod_{i=1, i \neq k}^{L} \frac{\bar{\gamma}_k}{\bar{\gamma}_k - \bar{\gamma}_i}$, provided $\bar{\gamma}_k \neq \bar{\gamma}_i$ for $k \neq i$.

$$
\begin{cases}\n\mathsf{BPSK}: \left\{\begin{array}{l} U_{1,\ell} \approx \sum_{k=1}^{L} \alpha_k^2 \mathbf{Re}\left[\left\langle \mathsf{s}_{1,\ell}\left(t\right), \tilde{\mathsf{s}}_{1,\ell}\left(t\right) \right\rangle\right] + \sum_{k=1}^{L} \alpha_k n_{k,\ell} \ U_{2,\ell} \approx \sum_{k=1}^{L} \alpha_k^2 \mathbf{Re}\left[\left\langle \mathsf{s}_{1,\ell}\left(t\right), \tilde{\mathsf{s}}_{2,\ell}\left(t\right) \right\rangle\right] + \sum_{k=1}^{L} \alpha_k n_{k,\ell} \ U_{1,\ell} \approx \sum_{k=1}^{L} \alpha_k^2 \mathbf{Re}\left[\left\langle \mathsf{s}_{1,\ell}\left(t\right), \tilde{\mathsf{s}}_{1,\ell}\left(t\right) \right\rangle\right] + \sum_{k=1}^{L} \alpha_k n_{k,\ell} \ U_{2,\ell} \approx \sum_{k=1}^{L} \alpha_k^2 \mathbf{Re}\left[\left\langle \mathsf{s}_{1,\ell}\left(t\right), \tilde{\mathsf{s}}_{2,\ell}\left(t\right) \right\rangle\right] + \sum_{k=1}^{L} \alpha_k n_{k,\ell}\ u_{2,\ell} \approx \sum_{k=1}^{L} \alpha_k^2 \mathbf{Re}\left[\left\langle \mathsf{s}_{1,\ell}\left(t\right), \tilde{\mathsf{s}}_{2,\ell}\left(t\right) \right\rangle\right] + \sum_{k=1}^{L} \alpha_k n_{k,\ell}\ u_{2,\ell} \end{array}\right.\n\end{cases}
$$

$$
\Rightarrow \begin{cases} \text{BPSK}: \left\{ \begin{array}{l} U_{1,\ell} \approx \sum_{k=1}^{L} \alpha_k^2 \sqrt{2\mathcal{E}_s} + \sum_{k=1}^{L} \alpha_k n_{k,\ell} \\ U_{2,\ell} \approx \sum_{k=1}^{L} \alpha_k^2 (-\sqrt{2\mathcal{E}_s}) + \sum_{k=1}^{L} \alpha_k n_{k,\ell} \\ U_{1,\ell} \approx \sum_{k=1}^{L} \alpha_k^2 \sqrt{2\mathcal{E}_s} + \sum_{k=1}^{L} \alpha_k n_{k,\ell} \end{array} \right. \\ \text{BFSK}: \left\{ \begin{array}{l} U_{1,\ell} \approx \sum_{k=1}^{L} \alpha_k^2 \sqrt{2\mathcal{E}_s} + \sum_{k=1}^{L} \alpha_k n_{k,\ell} \\ U_{2,\ell} \approx \sum_{k=1}^{L} \alpha_k^2 \cdot (0) + \sum_{k=1}^{L} \alpha_k n_{k,\ell} \end{array} \right. \end{cases}
$$

Then,

$$
P_e = \begin{cases} \frac{1}{2} \sum_{k=1}^{L} \pi_k \left(1 - \sqrt{\frac{\bar{\gamma}_k}{1 + \bar{\gamma}_k}} \right) \approx \binom{2L - 1}{L} \prod_{k=1}^{L} \frac{1}{4\bar{\gamma}_k}, & \text{BPSK, RAKE} \\ \frac{1}{2} \sum_{k=1}^{L} \pi_k \left(1 - \sqrt{\frac{\bar{\gamma}_k}{2 + \bar{\gamma}_k}} \right) \approx \binom{2L - 1}{L} \prod_{k=1}^{L} \frac{1}{2\bar{\gamma}_k}, & \text{BFSK, RAKE} \end{cases}
$$

Estimation of c_k

For orthogonal signaling, we can estimate c_n via

$$
\int_{0}^{T} r_{\ell} \left(t + \frac{n}{W} \right) \left(s_{1,\ell}^{*}(t) + \dots + s_{M,\ell}^{*}(t) \right) dt
$$
\n
$$
= \sum_{k=1}^{L} c_{k} \int_{0}^{T} s_{m,\ell} \left(t + \frac{n}{W} - \frac{k}{W} \right) \left(s_{1,\ell}^{*}(t) + \dots + s_{M,\ell}^{*}(t) \right) dt
$$
\n
$$
+ \int_{0}^{T} z \left(t + \frac{n}{W} \right) \left(s_{1,\ell}^{*}(t) + \dots + s_{M,\ell}^{*}(t) \right) dt
$$
\n
$$
= \sum_{k=1}^{L} c_{k} \int_{0}^{T} s_{m,\ell} \left(t + \frac{n}{W} - \frac{k}{W} \right) s_{m,\ell}^{*}(t) dt
$$
\n
$$
+ \int_{0}^{T} z \left(t + \frac{n}{W} \right) \left(s_{1,\ell}^{*}(t) + \dots + s_{M,\ell}^{*}(t) \right) dt \quad \text{(Orthogonality)}
$$
\n
$$
= c_{n} \int_{0}^{T} |s_{m,\ell}(t)|^{2} dt + \text{noise term} \quad \text{(Add-and-delay)}
$$

$M = 2$ case

Decision-feedback estimator

The previous estimator only works for orthogonal signaling. For, e.g., PAM signal with

$$
s_{\ell}(t) = I \cdot g(t) \text{ where } I \in \{\pm 1, \pm 3, \ldots, \pm (M-1)\},\
$$

we can estimate c_n via

$$
\int_0^T r_\ell \left(t + \frac{n}{W} \right) g^*(t) dt
$$
\n
$$
= \int_0^T \left(\sum_{k=1}^L c_k \cdot l \cdot g \left(t + \frac{n}{W} - \frac{k}{W} \right) + z \left(t + \frac{n}{W} \right) \right) g^*(t) dt
$$
\n
$$
= \sum_{k=1}^L c_k \cdot l \cdot \int_0^T g \left(t + \frac{n}{W} - \frac{k}{W} \right) g^*(t) dt + \text{noise term}
$$
\n
$$
= c_n \cdot l \cdot \int_0^T |g(t)|^2 dt + \text{noise term} \quad \text{(Add-and-delay)}
$$

Usually it requires $\frac{(\Delta t)_c}{T} > 100$ in order to have an accurate estimate of $\{c_n\}_{n=1}^L$.

Note that for DPSK and FSK with square-law combiner, it is unnecessary to estimate $\{c_n\}_{n=1}^L$.

So, they have no further performance loss (due to an inaccurate estimate of $\{c_n\}_{n=1}^L$).

What you learn from Chapter 13

- Statistical model of (WSSUS) (linear) multipath fading channels:
	- $c_{\ell}(\tau; t) = c(\tau; t)e^{-i2\pi f_c \tau}$ and $c(\tau; t) = |c_{\ell}(\tau; t)|$
	- Multipath intensity profile or delay power spectrum

$$
R_{c_{\ell}}(\tau) = R_{c_{\ell}}(\tau; \Delta t = 0).
$$

- Multipath delay spread T_m vs coherent bandwidth $(\Delta f)_c$
- Frequency-selective vs frequency-nonselective
- Spaced-frequency, spaced-time correlation function

$$
R_{\mathbf{C}_{\ell}}(\Delta f; \Delta t) = \mathbb{E}\left\{ \mathbf{C}_{\ell}(f + \Delta f; t + \Delta t)\mathbf{C}_{\ell}^{*}(f; t) \right\}
$$

• Doppler power spectrum

$$
S_{\mathbf{C}_{\ell}}(\lambda) = \int_{-\infty}^{\infty} R_{\mathbf{C}_{\ell}}(\Delta f = 0; \Delta t) e^{-i 2\pi \lambda (\Delta t)} d(\Delta t)
$$

- Doppler spread B_d vs coherent time $(\Delta t)_c$
- Slow fading versus fast fading
- Scattering function

$$
S(\tau;\lambda)=\mathcal{F}_{\Delta t}\left\{R_{c_{\ell}}(\tau;\Delta t)\right\}
$$

- Jakes' model
- Rayleigh, Rice and Nakagami-m, Rummler's 3-path model
- Deep fading phenomenon
- B_dT_m spread factor: Underspread vs overspread
- Analysis of error rate under frequency-nonselective, slowly Rayleigh- and Nakagami-m-distributed fading channels $($ ≡diversity under Rayleigh) with $M = 2$
	- (Good to know) Analysis of the error rate \cdots with $M > 2$.
- Rake receiver under frequency-selective, slowly fading channels
	- Assumption: Bandlimited signal with ideal lowpass filter and perfect channel estimator at the receiver
	- This assumption results in a (finite-length) tapped-delay-line channel model under a finite delay spread.
	- Error analysis under add-and-delay assumption on the transmitted signals