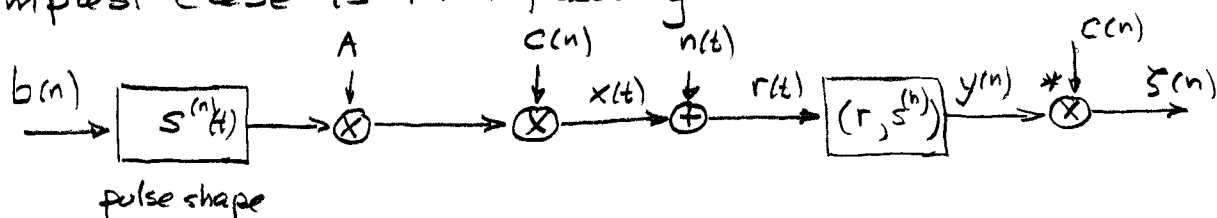


3.2 Effect of Fading on Single-User BER

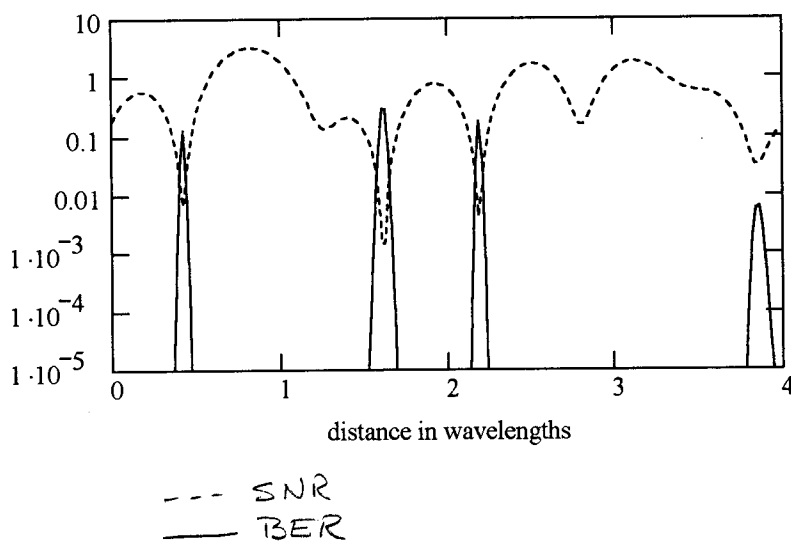
- Fading has a devastating effect on BER, as we'll see below.
- Simplest case is flat fading.



This model assumes Rx has perfect channel/state information (CSI). For given $c(n)$, the signal component of $z(n)$ is $|c(n)|^2 A b$ and the variance of the noise component is $N_0 |c|^2$. Hence BER is (BPSK)

$$P_e(c) = Q\left(\frac{|c|^2 A}{\sqrt{N_0 |c|^2}}\right) = Q\left(\sqrt{\frac{|c|^2 2E_b}{N_0}}\right) = Q\left(\sqrt{2\Gamma_b |c|^2}\right)$$

- Typical variation with time:



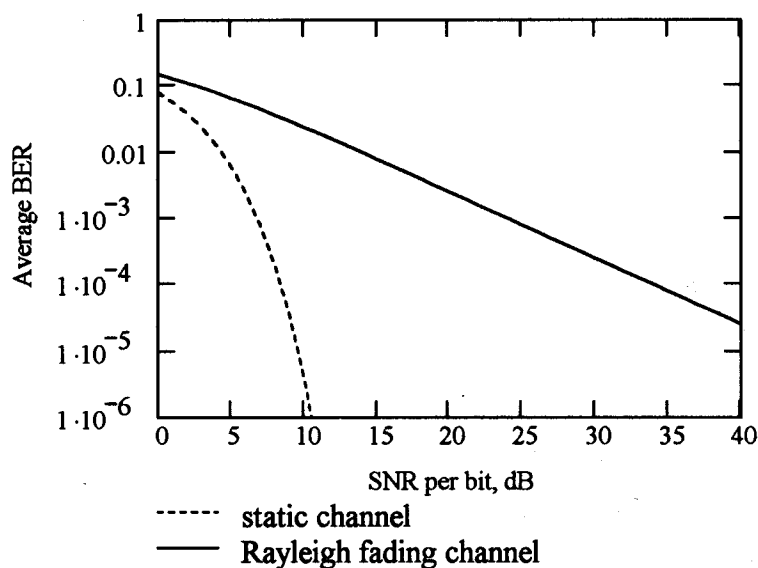
- bursty
- prob of deep fades determines the average BER

- To obtain the BER, average the instantaneous BER with the pdf of channel gain. Assume $\sigma_c^2 = \frac{1}{2}$, so $\overline{|c|^2} = \overline{z} = 1$, which leaves the average received SNR equal to average transmitted SNR.

$$\overline{P_e} = \int_0^{\infty} e^{-z} Q(\sqrt{2\Gamma_b z}) dz \quad [\text{Proa 95, Stub 96, Schw 66}]$$

$$= \frac{1}{2} \left(1 - \sqrt{\frac{\Gamma_b}{1+\Gamma_b}} \right) \quad (\text{integration by parts})$$

$$\sim 1/4\Gamma_b \quad \text{large SNR}$$



- A 3 dB increase in SNR roughly
 - squares BER in static
 - halves BER in fading

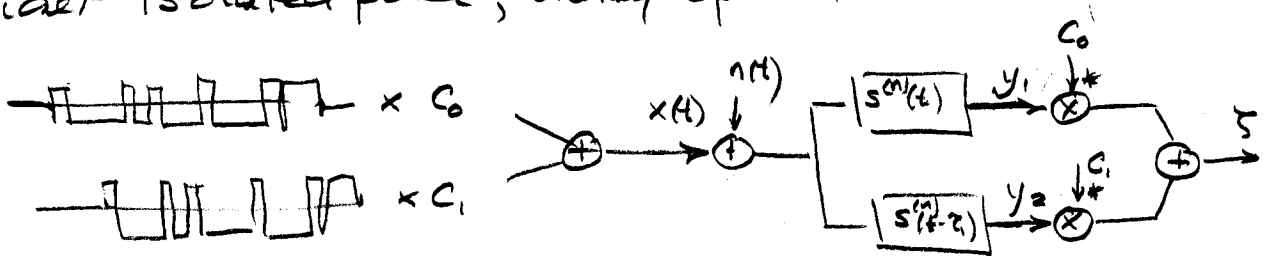
- What if CSI is not perfect?

$$\hat{c}(n) = Cc(n) + e(n)$$

Very difficult integrals — but there's an easy way using Gaussian quadratic forms, as we'll see.

- Next step: selective fading. We won't do the full solution at this point. Objectives are just:
 - demonstrate that sufficient stat ξ , the Rake receiver, follows from ML principles, too.
 - demonstrate a form of diversity reception.

• Consider isolated pulse, delay spread



- From pp 2.3.7, 2.3.8, sufficient stat is

$$\xi = \xi^T R \xi + A b + \xi^T v = \xi^T y$$

$$R = \begin{bmatrix} 1 & r^{(n)}(t-\tau) \\ r^{(n)}(\tau) & 1 \end{bmatrix}, \quad R_v = \frac{1}{2} [v v^T] = N_0 R$$

R is Hermitian, positive definite, so

$\xi^T R \xi + A$ is pos, real \Rightarrow slice ξ for decision

- This combining also follows from ML principles.

The conditional prob is

$$P_{y|b}(y|b) = \frac{1}{(2\pi)^2 |R|} \exp\left(-\frac{1}{2} (y - R \underline{c} A b)^{\dagger} R^{-1} (y - R \underline{c} A b)\right)$$

and we max wrt b . Log likelihood, keep exponent

$$\hat{b} = \arg \min \left[(y - R \underline{c} A b)^{\dagger} R^{-1} (y - R \underline{c} A b) \right]$$

$$= \arg \min \left[y^{\dagger} R^{-1} y - 2 b A \operatorname{Re} \left[\underline{c}^{\dagger} R R^{-1} y \right] + b A \underbrace{\underline{c}^{\dagger} R R^{-1} R \underline{c}}_R A b \right]$$

$$= \arg \max \left[b A \operatorname{Re} \left[\underline{c}^{\dagger} y \right] \right]$$

$$= \operatorname{sgn}(\operatorname{Re}[\underline{c}])$$

So we have another justification of this combining.

- Next, the BER:

$$\underline{y} = \underline{c}^{\dagger} \begin{bmatrix} 1 & r^{(n)}(z) \\ r^{(n)}(z) & 1 \end{bmatrix} \underline{c} A b + \eta \quad \sigma_{\eta}^2 = N_0 \underline{c}^{\dagger} R \underline{c}$$

The conditional prob is easy. Average over \underline{c} ?

$$\underline{c}^{\dagger} R \underline{c} = |c_0|^2 + |c_1|^2 + 2 \operatorname{Re} [c_0 c_1^* r^{(n)}(z)]$$

The cross term makes it tough - again, need theory of Gaussian quad forms.

However, if $|r^{(n)}(\tau)| \ll 1$, as in spread spectrum,

$$S \approx (|k_0|^2 + |k_1|^2) A b + \eta \quad \sigma_\eta^2 = N_0 (|k_0|^2 + |k_1|^2)$$

we can see the diversity effect:

the prob that both $|k_0|^2$ and $|k_1|^2$ are very small is much less than the individual probabilities of being small, so the multipath should improve the BER

- Postpone further analysis until we have the mathematical tools.