

# *Coded Multiband OFDM with Tone Interference*

*or, “How to Analyze Coded OFDM Error Rates in Quasi-Static Channels”*

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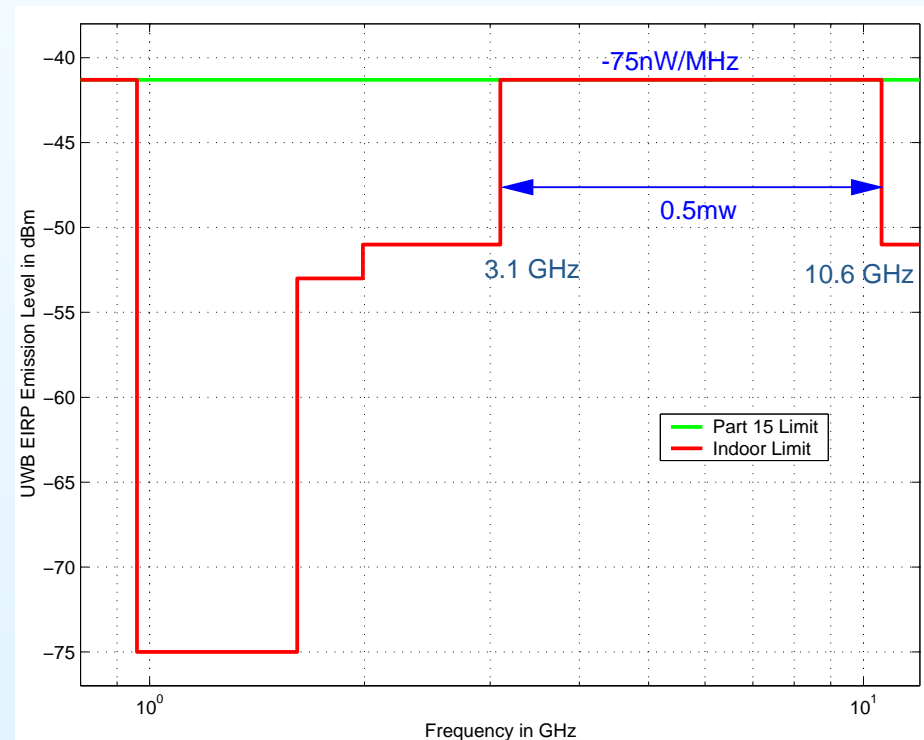
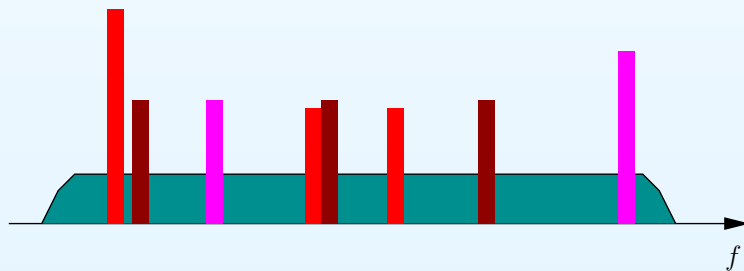
# Agenda

- What is UWB?
- Multiband OFDM proposal for high data-rate UWB
- Performance Analysis of coded OFDM systems
  - Per-realization BER analysis
  - Direct average BER analysis
- Performance results for Multiband OFDM
  - without interference
  - with interference
- Conclusions



# Ultra-Wideband (UWB) Wireless Transmission

- UWB is a wireless technology that uses an ultra-wide bandwidth ( $> 500$  MHz) and operates at very low power
- Underlay (reuse spectrum, potential for interference)
- FCC: license-exempt operation in 3.1–10.6 GHz band



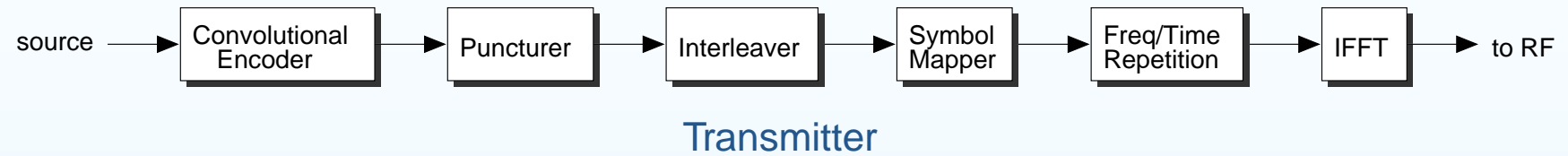
## Multiband OFDM: Overview

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- Multiband OFDM is a leading proposal for high-rate UWB communication
- Multiband OFDM Alliance (MBOA) formed in 2003, now with over 170 member companies (Intel, Texas Instruments, . . .)
- Standardization of Multiband OFDM by
  - ~~IEEE TG 802.15.3a~~ (high-rate WPANs)
  - ECMA-368 High Rate Ultra Wideband PHY and MAC Standard (December 2005)
- Will be used in (for example)
  - Wireless USB
  - Wireless 1394 (Firewire)



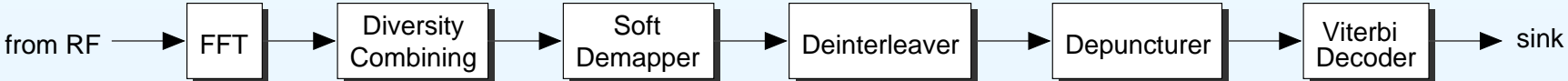
# Multiband OFDM: System Model



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Transmitter



Receiver



## Multiband OFDM: 802.15.3a Channel Model

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- Impulse response consists of clusters of multipath components
- The channel is **quasi-static**
- The channel stays constant for (at least) one packet transmission
- **We can only code over one realization of the channel**
- (Also turns out the frequency-domain channel gains are Rayleigh — will be important later)



# Performance Measures

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2. Outage BER
  - if we assume  $X\%$  of channels are bad (“in outage”), what is the worst-case performance of the remaining  $(100 - X)\%$  of the channels?
  - (we will consider  $X = 10\%$  outage)



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Outage BER is a more relevant measure for quasi-static channels, since the channel may be constant for a long time...

**BUT, it is hard to determine.**



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What is the “standard” method for analyzing coded modulation?



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- $k/n$  convolutional code (rate  $R_c$ )
- $d_{\text{free}}$  = free distance
- $\beta_d$  = total number of bit errors associated with paths of Hamming weight  $d$



## Classical Average BER Analysis

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- $k/n$  convolutional code (rate  $R_c$ )
- $d_{\text{free}}$  = free distance
- $\beta_d$  = total number of bit errors associated with paths of Hamming weight  $d$
- Average BER is given by the union bound

$$\text{BER} \leq \frac{1}{k} \sum_{d=d_{\text{free}}}^{\infty} \beta_d P_2(d)$$

where  $P_2(d)$  is the pairwise error probability (PEP) of two sequences differing in  $d$  positions

- Can get expressions for  $P_2(d)$  in different types of channels (AWGN, fading, etc)



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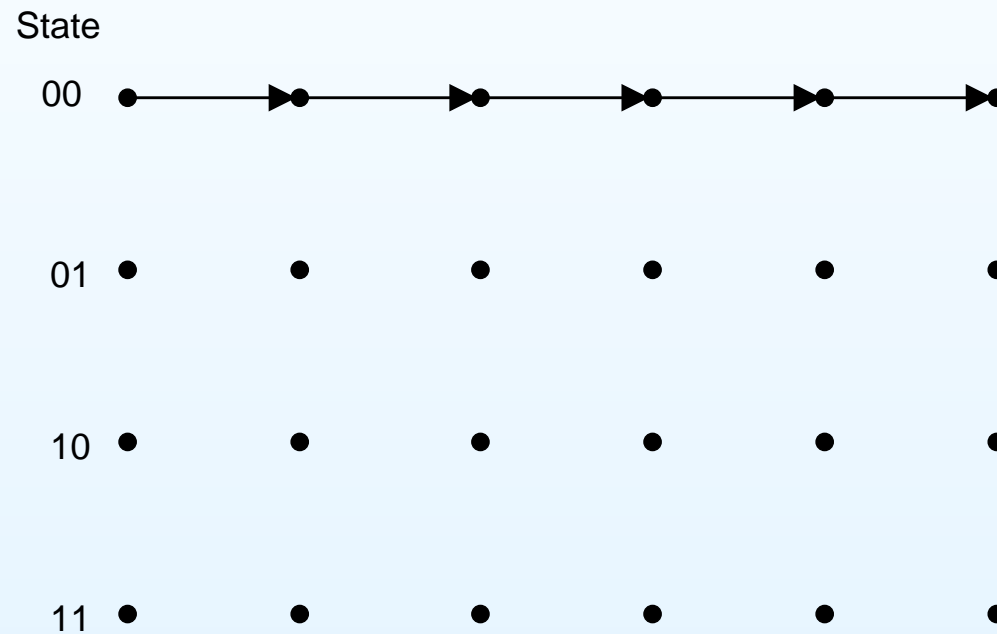
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- Correlation between bits depends on starting position
- **Integrating over the pdf of channel distribution is NOT the thing to do here**
- Question: if we know channel gains, can we predict the BER?
- Answer: Yes. Let’s examine the trellis of the code...



# Error Vectors

Trellis representation of  $R_c = 1/2 (7,5)$  convolutional code

An example error path:



Input: { , , , }

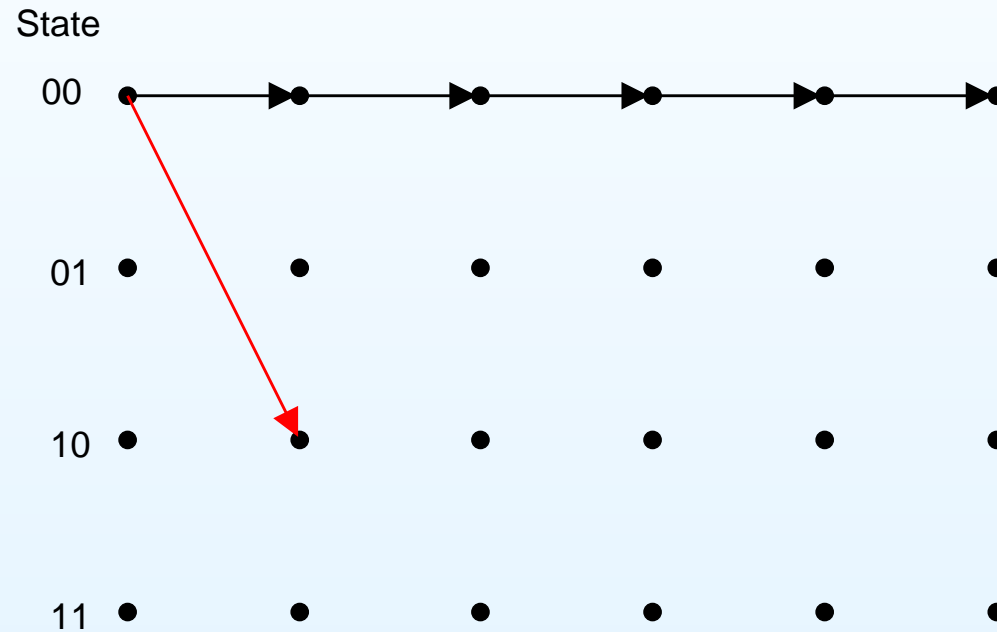
Output  $e_1$ : { , , , , , , , }



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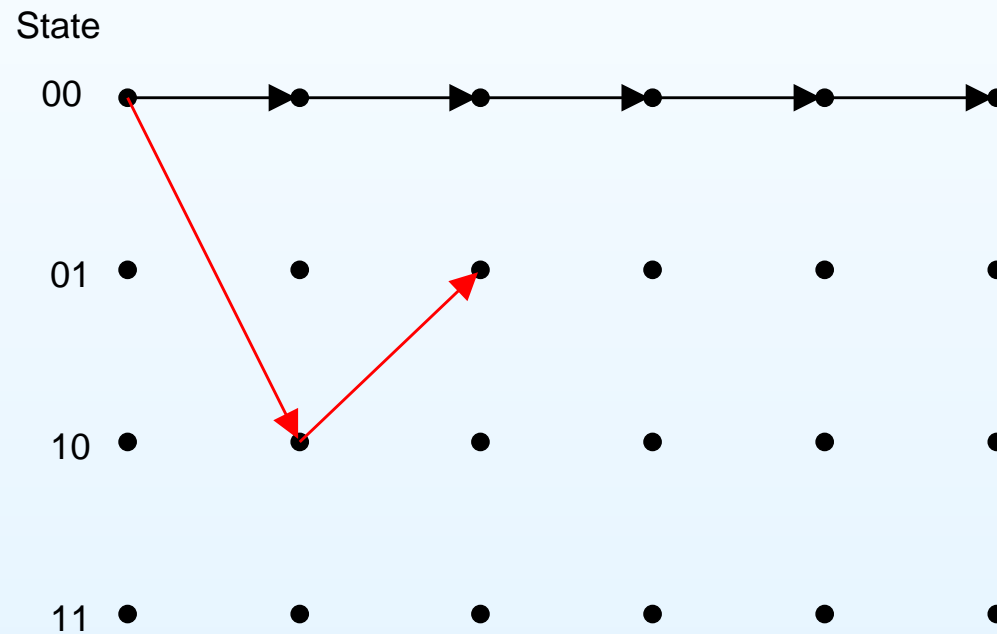
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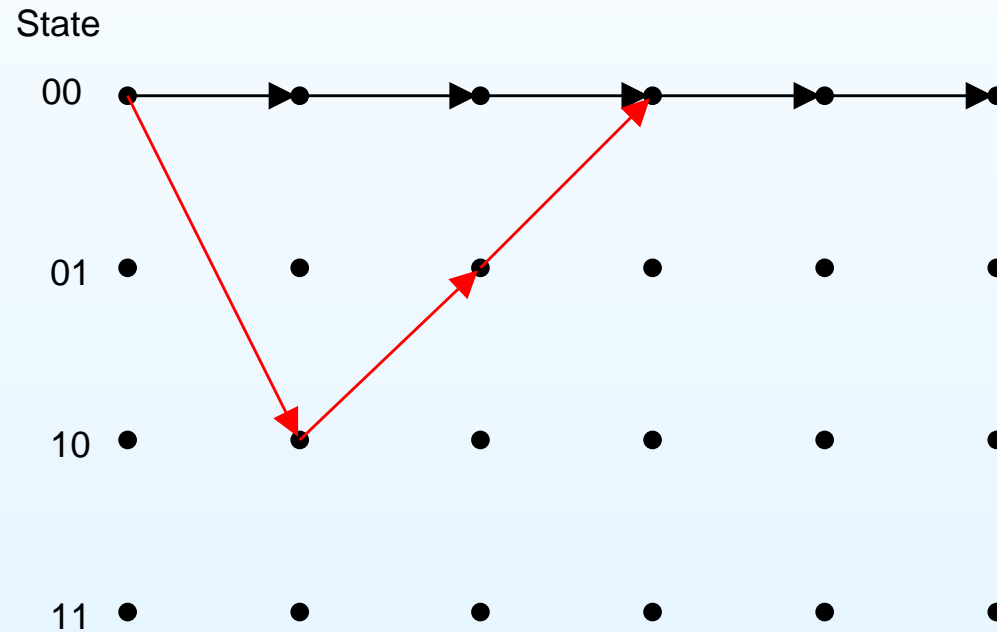
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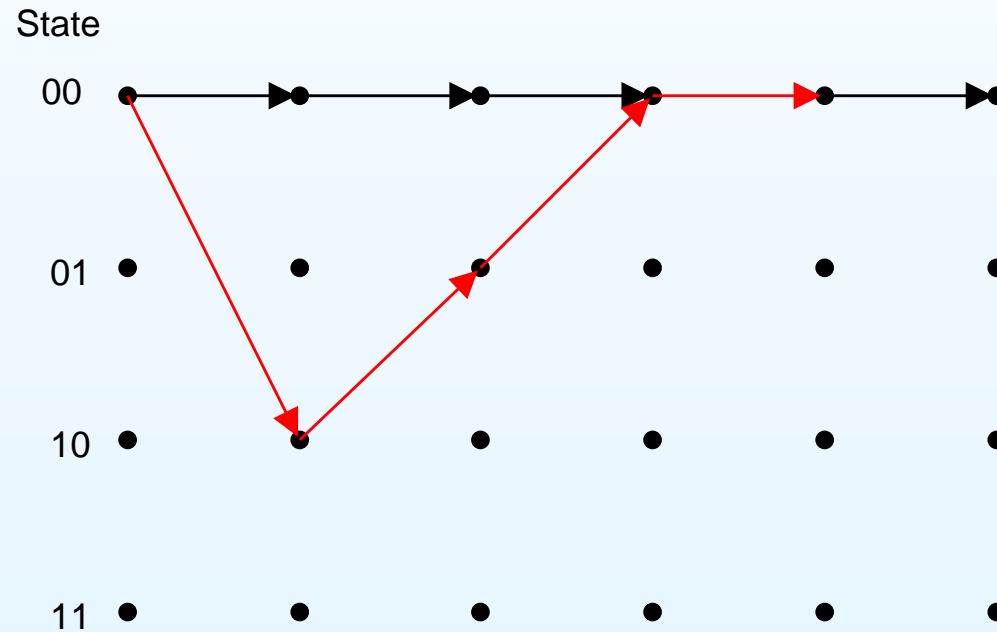
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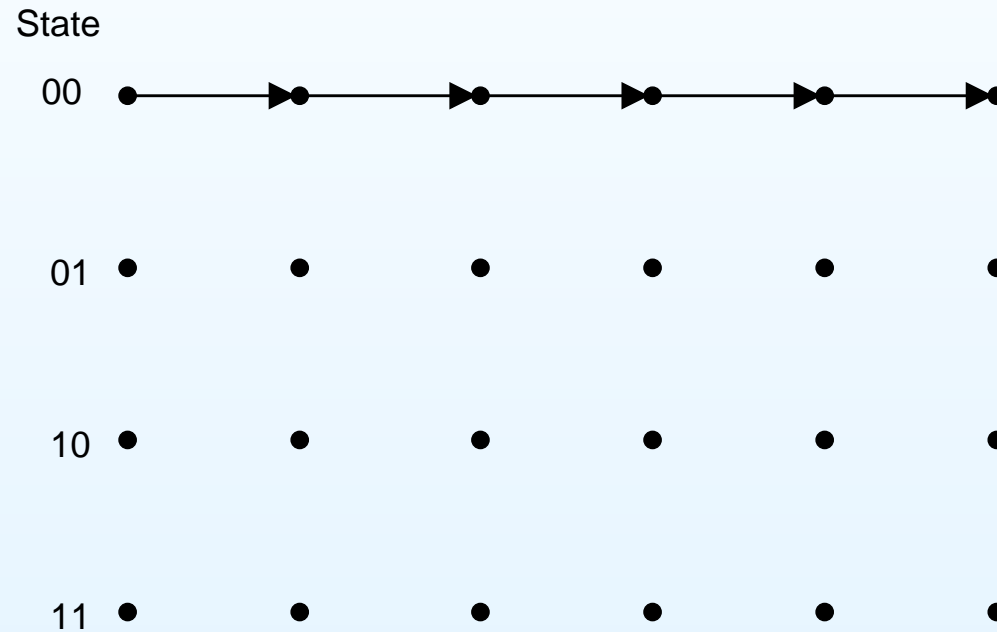
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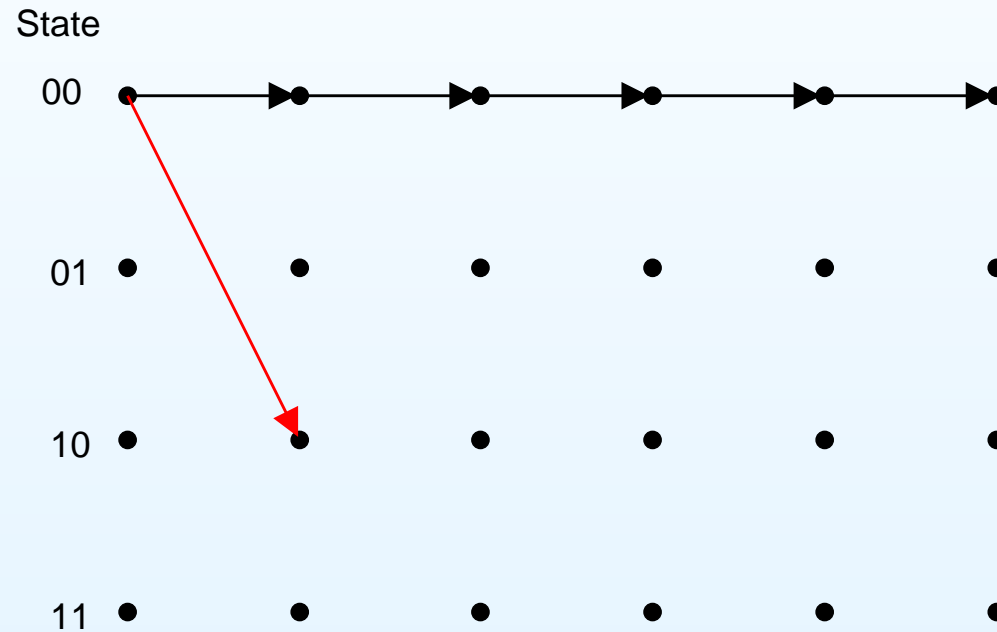
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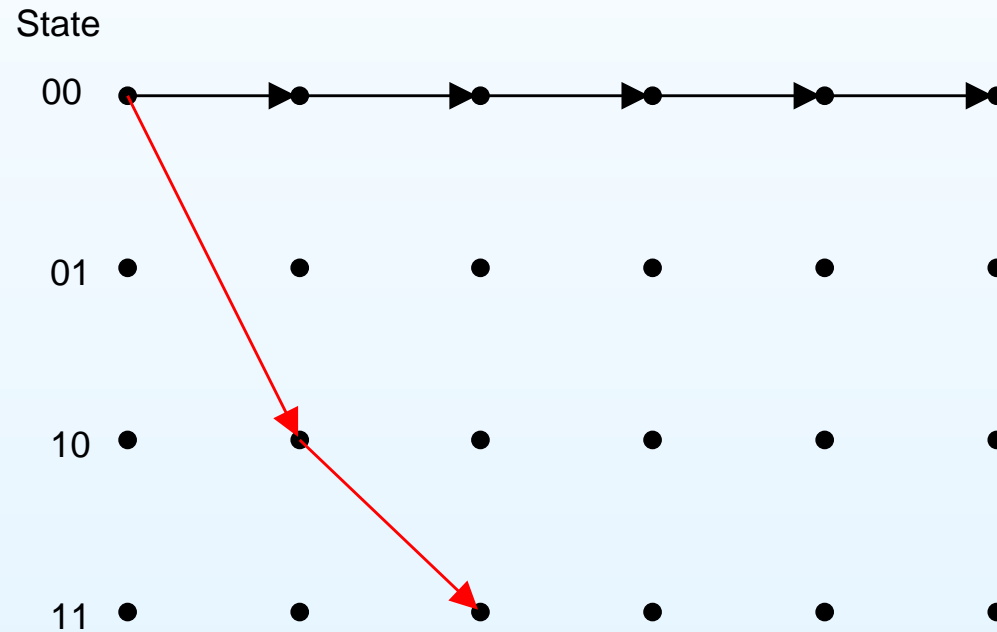
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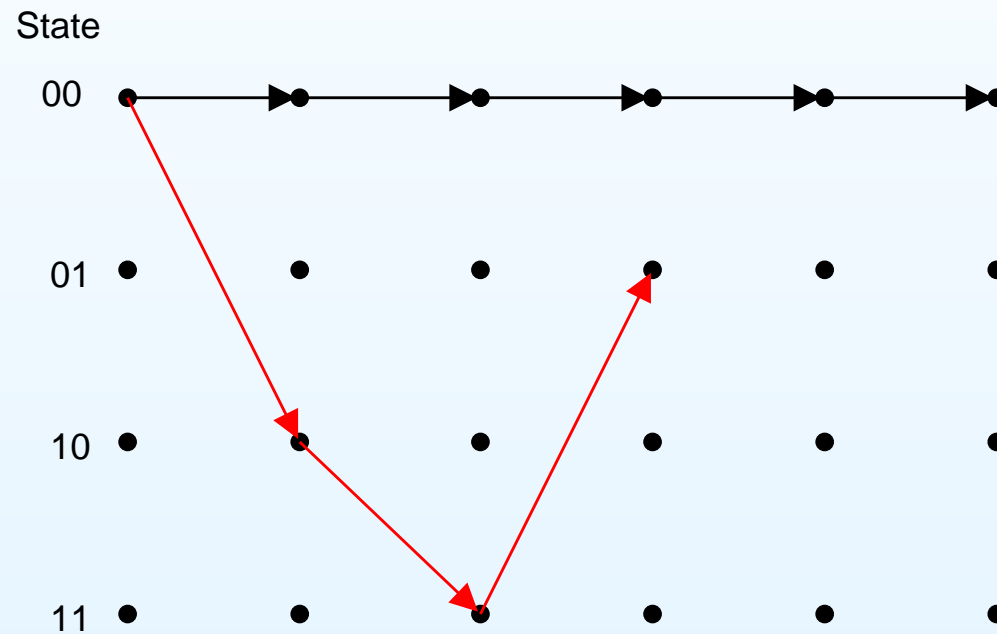
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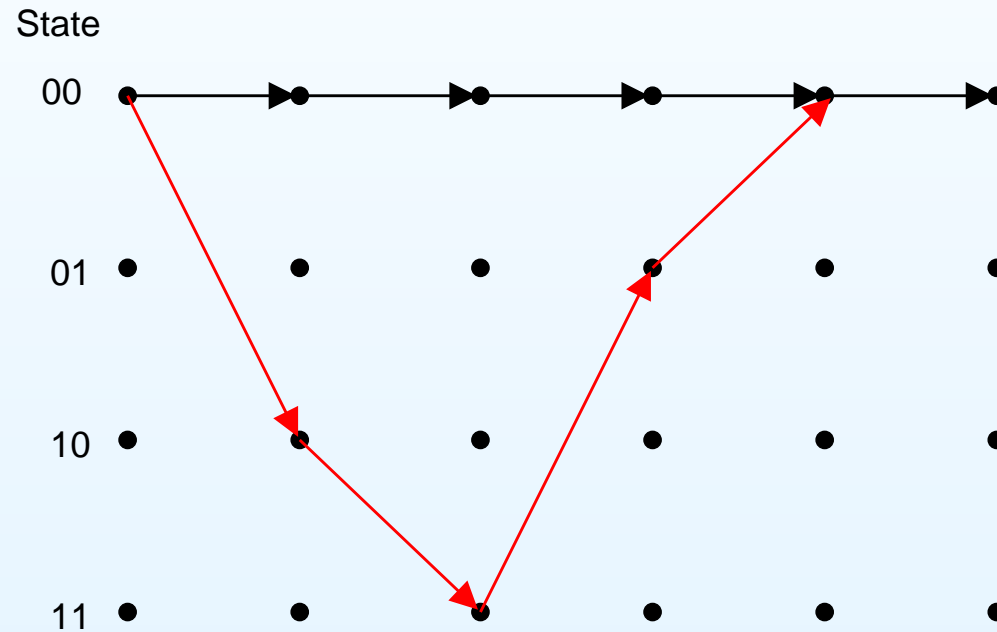
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$\mathcal{E}$  contains all the “most likely” error events (the ones with shortest length)



## System Model

We consider the following frequency-domain system model

$$r = Hx + J + n$$

where

- $x$  are the transmitted symbols
- $H = \text{diag}(h)$  is a (diagonal) matrix of channel gains
- $J$  is the interference (frequency-domain)
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Interference is modelled as

$$i(t) = \sum_{k=1}^{N_i} \alpha_k e^{j(2\pi f_k t + \phi_k)}$$



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- For each error vector  $e_\ell$ , we form the “full” **error codeword** (length  $L_c$ )

$$\mathbf{q}_{i,\ell} = \left[ \underbrace{0 \ 0 \ \dots \ 0}_{i-1} \ \underbrace{\mathbf{e}_\ell}_{l_\ell} \ \underbrace{0 \ 0 \ \dots \ 0}_{L_c - l_\ell - i + 1} \right]^T$$

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- Now, given  $\mathbf{q}_{i,\ell}$ , the **competing codeword** is given by

$$\mathbf{v}_{i,\ell} = \mathbf{c} \oplus \mathbf{q}_{i,\ell}$$



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- true codeword  $c \longrightarrow x$
- competing codeword  $v_{i,l} \longrightarrow z_{i,l}$
- the PEP for the  $l$ th error vector starting in the  $i$ th position, i.e., the probability that  $v_{i,l}$  is detected given that  $c$  was transmitted, is given by

$$\text{PEP}_{i,l}(\mathbf{H}, \mathbf{J}) = \Pr \{ \|\mathbf{r} - \mathbf{H}\mathbf{x}\|^2 > \|\mathbf{r} - \mathbf{H}\mathbf{z}_{i,l}\|^2 \mid \mathbf{H}, \mathbf{J} \}$$



## Method I: Per-realization BER

---

For given  $\mathbf{H}$ ,  $\mathbf{J}$  we can write

$$\text{PEP}_{i,l}(\mathbf{H}, \mathbf{J}) = Q \left( \frac{\frac{1}{2} \|\mathbf{H}(\mathbf{x} - \mathbf{z}_{i,l})\|^2 + \text{Re} \{ \mathbf{J}^H \mathbf{H}(\mathbf{x} - \mathbf{z}_{i,l}) \}}{\sqrt{\frac{1}{2} \mathcal{N}_0 \|\mathbf{H}(\mathbf{x} - \mathbf{z}_{i,l})\|^2}} \right)$$

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and if  $\mathbf{J} = \mathbf{0}$ , we can get

$$\text{PEP}_{i,l}(\mathbf{H}, \mathbf{J}) = Q \left( \sqrt{\frac{\|\mathbf{H}(\mathbf{x} - \mathbf{z}_{i,l})\|^2}{2\mathcal{N}_0}} \right)$$



## Method I: Per-realization BER

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BER for  $e_\ell$ , starting in the  $i$ th position, is given by

$$P_{i,\ell}(\mathbf{H}, \mathbf{J}) = a_\ell \cdot \text{PEP}_{i,\ell}(\mathbf{H}, \mathbf{J})$$



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Sum over all  $e_\ell \longrightarrow$  BER for the  $i$ th position

$$P_i(\mathbf{H}, \mathbf{J}) = \sum_{\ell=1}^L P_{i,\ell}(\mathbf{H}, \mathbf{J})$$



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All positions equally likely  $\longrightarrow$  BER  $P(\mathbf{H}, \mathbf{J})$  is

$$P(\mathbf{H}, \mathbf{J}) = \frac{1}{L_c} \sum_{i=1}^{L_c} \min \left[ \frac{1}{2}, \sum_{\ell=1}^L P_{i,\ell}(\mathbf{H}, \mathbf{J}) \right]$$

This is the (approximate) BER for a given  $\mathbf{H}$



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Once we know  $P(\mathbf{H}, \mathbf{J})$ , we can easily obtain both average and outage BER



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Outage BER:

$$P_{\text{out}} = \max_{\mathbf{H}_i \in \mathcal{H}_{\text{in}}} \mathbb{E}_{\mathbf{J}} \{P(\mathbf{H}_i, \mathbf{J})\}$$

(where  $\mathcal{H}_{\text{in}}$  are the channels that are NOT in outage)



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- Able to obtain outage BER
- For average BER — slow!

To obtain average BER, we can do better for certain channel distributions

## Method II: Direct Average BER

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If channel gains  $h$  are

- Rayleigh-distributed r.v.s, and
- Correlated with correlation matrix  $\Sigma_{hh}$

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Start from PEP for position  $i$ , error vector  $\ell$

$$\text{PEP}_{i,\ell} = \Pr \{ \|\mathbf{r} - \mathbf{H}\mathbf{x}\|^2 > \|\mathbf{r} - \mathbf{H}\mathbf{z}_{i,\ell}\|^2 \mid \mathbf{H}, \mathbf{J} \}$$



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We only need to consider non-zero entries of  $(\mathbf{x} - \mathbf{z}_{i,\ell}) \longrightarrow$  make

- $\mathbf{x}'$ ,  $\mathbf{z}'_{i,\ell}$ ,  $\mathbf{H}' = \text{diag}(\mathbf{h}')$ ,  $\mathbf{J}'$ , and  $n'$
- $\Sigma_{h'h'}$
- $\mathbf{D} = \text{diag}(\mathbf{x}' - \mathbf{z}'_{i,\ell})$
- $\mathbf{g} = \mathbf{H}'(\mathbf{x}' - \mathbf{z}'_{i,\ell}) = \mathbf{D}\mathbf{h}'$



## Method II: Direct Average BER

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We have

$$\begin{aligned}\mathbb{E}(\mathbf{g}) &= \mathbf{0}_{\eta \times 1} \\ \mathbb{E}(\mathbf{g}\mathbf{g}^H) &= \mathbf{R}_{gg} = \mathbf{D}\boldsymbol{\Sigma}_{h'h'}\mathbf{D}^H\end{aligned}$$

i.e., the distribution of  $\mathbf{g}$  is zero-mean complex Gaussian with covariance matrix  $\mathbf{R}_{gg}$ .

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Rewriting the PEP to include only contributing terms, we obtain

$$\begin{aligned}\overline{\text{PEP}}_{i,l} &= \Pr \{ \|\mathbf{r}' - \mathbf{H}'\mathbf{z}'_{i,l}\|^2 - \|\mathbf{r}' - \mathbf{H}'\mathbf{x}'\|^2 < 0 \} \\ &= \Pr \{ \mathbf{g}\mathbf{g}^H - \mathbf{g}(\mathbf{J}' + \mathbf{n}')^H - (\mathbf{J}' + \mathbf{n}')\mathbf{g}^H < 0 \} \\ &= \Pr \{ \Delta_{i,l}(\mathbf{D}) < 0 \}\end{aligned}$$



## Method II: Direct Average BER

---

$$\Delta_{i,\ell}(\mathbf{D}) = \mathbf{y}^H \mathbf{A} \mathbf{y} \text{ and}$$

$$\mathbf{y} = \begin{bmatrix} \mathbf{g} \\ \mathbf{J}' + \mathbf{n}' \end{bmatrix}$$

$$\mathbf{A} = \begin{bmatrix} \mathbf{I}_\eta & -\mathbf{I}_\eta \\ -\mathbf{I}_\eta & \mathbf{0}_\eta \end{bmatrix}$$

and  $\mathbf{y}$  has

- mean  $\mu_{\mathbf{y}\mathbf{y}}$
- covariance matrix  $\mathbf{R}_{\mathbf{y}\mathbf{y}}$

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Laplace transform approach to get  $\Pr \{ \Delta_{i,\ell}(\mathbf{D}) < 0 \}$



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- non-faded interferers

$$\Phi_{i,l}(s) = \frac{\exp[-s\boldsymbol{\mu}_{yy}^H (\mathbf{A}^{-1} + s\mathbf{R}_{yy})^{-1} \boldsymbol{\mu}_{yy}]}{\det(\mathbf{I}_{2\eta} + s\mathbf{R}_{yy}\mathbf{A})}$$

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$$\Phi_{i,l}(s) = \frac{\exp[-s\boldsymbol{\mu}_{yy}^H(\mathbf{A}^{-1} + s\mathbf{R}_{yy})^{-1}\boldsymbol{\mu}_{yy}]}{\det(\mathbf{I}_{2\eta} + s\mathbf{R}_{yy}\mathbf{A})}$$

- iid Rayleigh interferers

$$\Phi_{i,l}(s) = \frac{1}{\det(\mathbf{I}_{2\eta} + s\mathbf{R}_{yy}\mathbf{A})}$$

## Method II: Laplace Transforms

Laplace transforms of  $\Delta_{i,l}(\mathbf{D})$  are

- non-faded interferers

$$\Phi_{i,l}(s) = \frac{\exp[-s\boldsymbol{\mu}_{\mathbf{y}\mathbf{y}}^H(\mathbf{A}^{-1} + s\mathbf{R}_{\mathbf{y}\mathbf{y}})^{-1}\boldsymbol{\mu}_{\mathbf{y}\mathbf{y}}]}{\det(\mathbf{I}_{2\eta} + s\mathbf{R}_{\mathbf{y}\mathbf{y}}\mathbf{A})}$$

- iid Rayleigh interferers

$$\Phi_{i,l}(s) = \frac{1}{\det(\mathbf{I}_{2\eta} + s\mathbf{R}_{\mathbf{y}\mathbf{y}}\mathbf{A})}$$

and

$$\overline{\text{PEP}}_{i,l} = \Pr\{\Delta_{i,l}(\mathbf{D}) < 0\} = \frac{1}{2\pi j} \int_{c-j\infty}^{c+j\infty} \Phi_{i,l}(s) \frac{ds}{s}$$

## Method II: Average BER

---

$\overline{\text{BER}}$  for  $e_\ell$ , starting in the  $i$ th position, is given by

$$\bar{P}_{i,\ell} = a_\ell \cdot \overline{\text{PEP}}_{i,\ell}$$

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Sum over all  $e_\ell \longrightarrow$  BER for the  $i$ th position

$$\bar{P}_i = \sum_{\ell=1}^L \bar{P}_{i,\ell}$$



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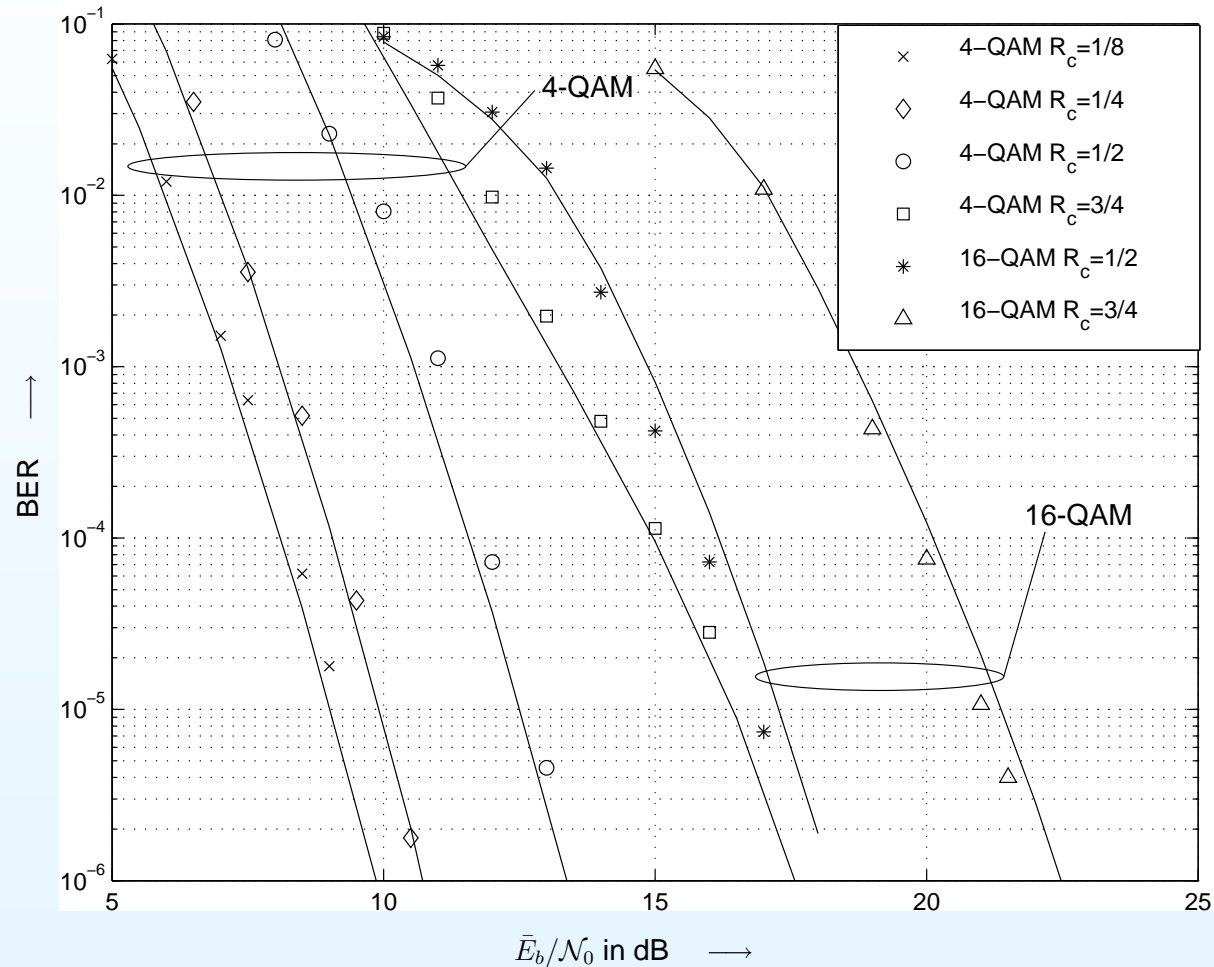
All positions equally likely  $\longrightarrow$   $\overline{\text{BER}}$  is

$$\bar{P} = \frac{1}{L_c} \sum_{i=1}^{L_c} \bar{P}_i = \frac{1}{L_c} \sum_{i=1}^{L_c} \sum_{\ell=1}^L \bar{P}_{i,\ell}$$

This is the (approximate) average BER



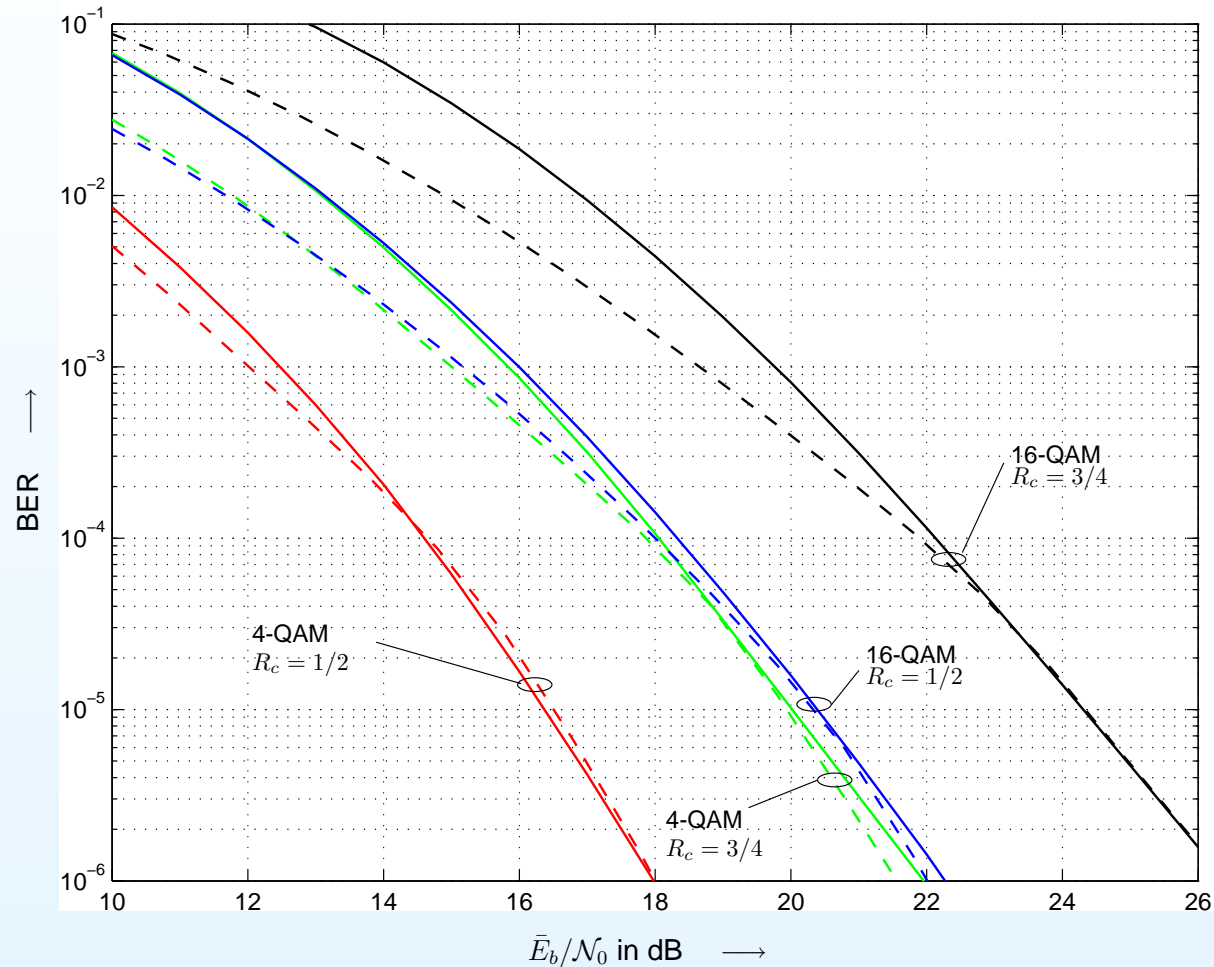
# Results: Outage BER, no interference



10% outage BER vs.  $\bar{E}_b/\mathcal{N}_0$  from Method I (lines) and simulation results (markers) for different code rates and modulation schemes. UWB CM1 channel. Code rates 1/4 and 1/8 include repetition. No interference ( $\mathbf{J} = \mathbf{0}_{N \times 1}$ ).



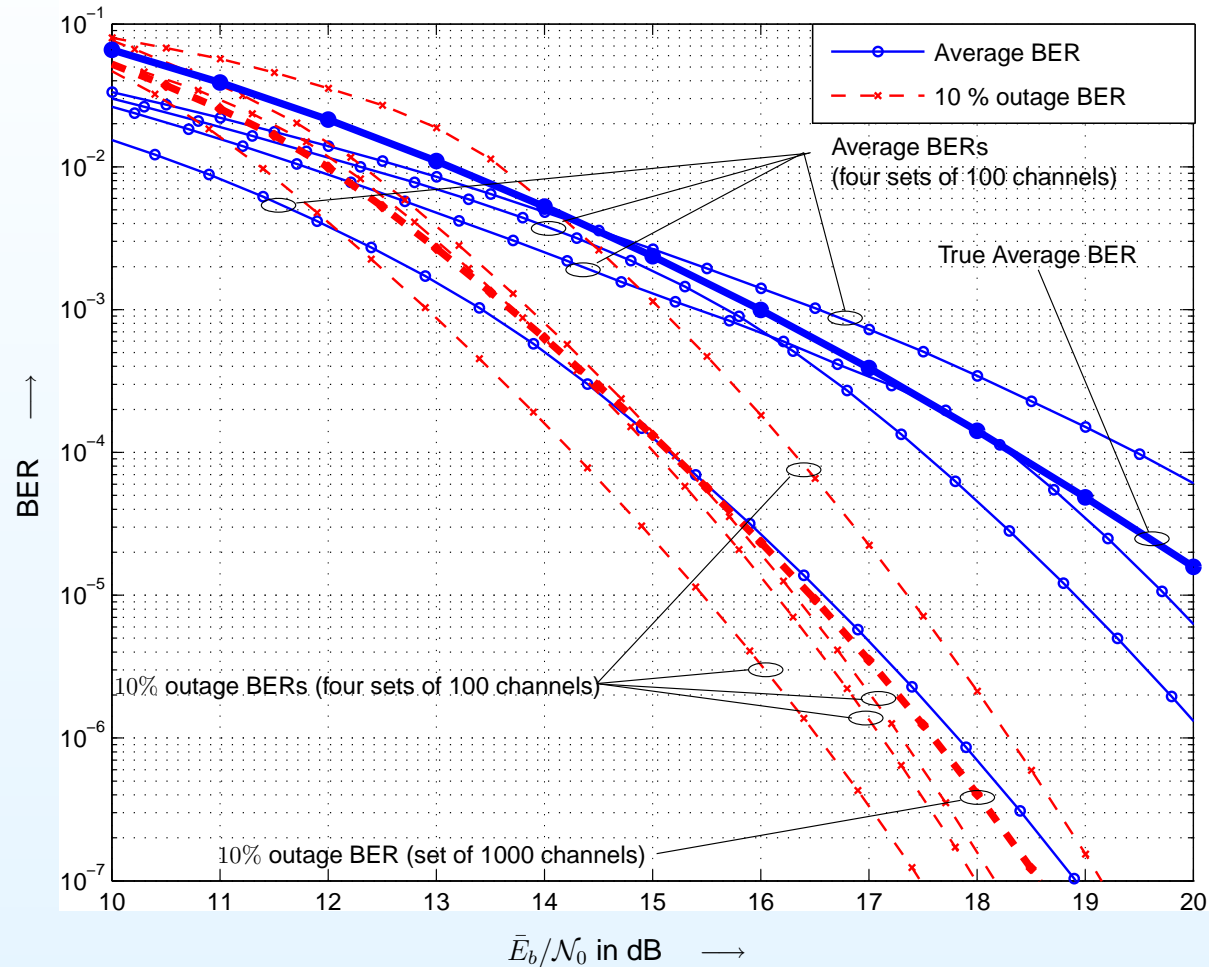
# Results: Average BER, no interference



Average BER versus  $\bar{E}_b/\mathcal{N}_0$  for 4-QAM and 16-QAM with code rates  $R_c = 1/2$  and  $3/4$ . Solid lines: Direct average from Method II. Dashed lines: Method I with an average over 10,000 channel realizations. UWB CM1 channel. No interference ( $\mathbf{J} = \mathbf{0}_{N \times 1}$ ).



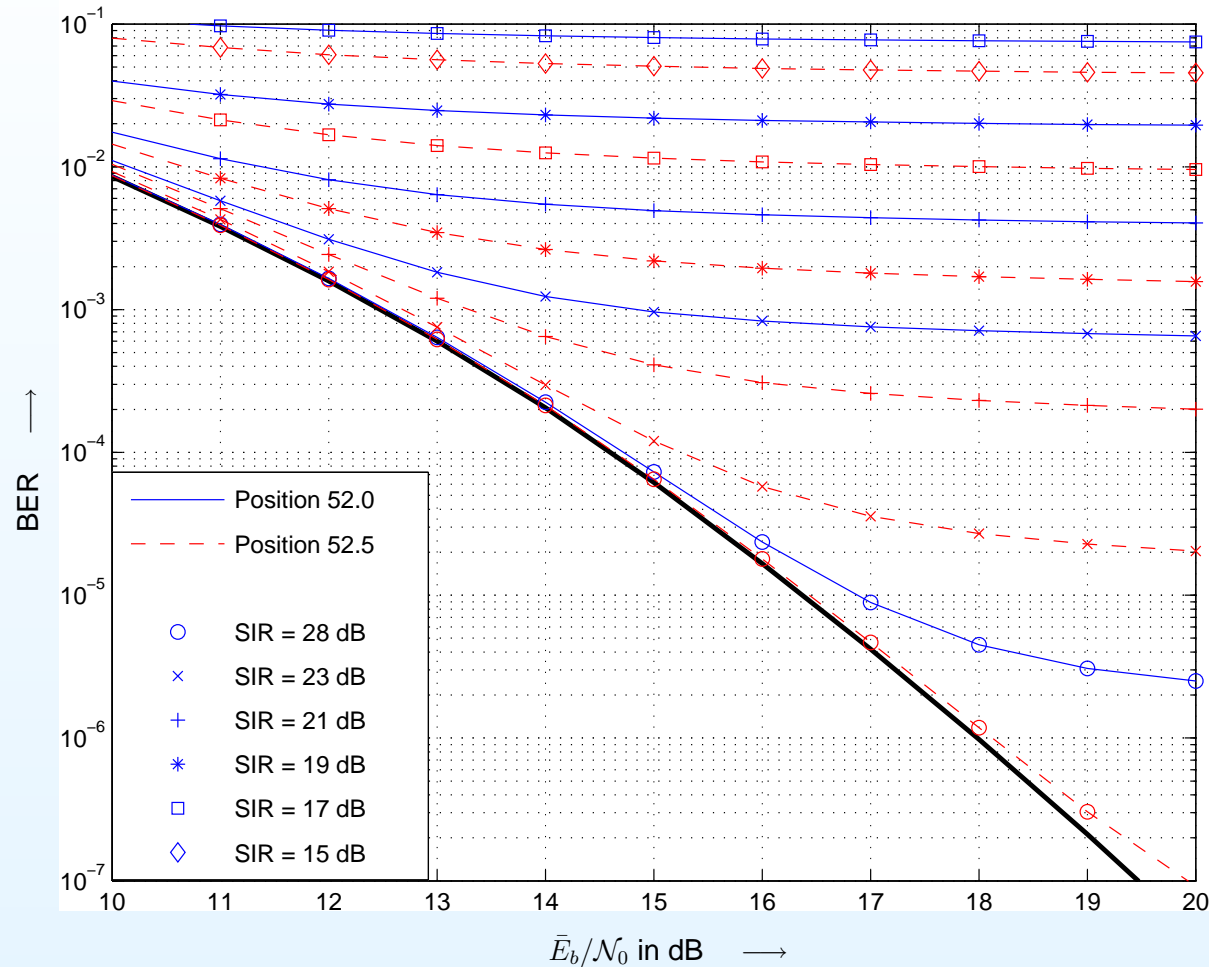
# Results: Variation in Realizations



Average BER (solid lines) and 10% outage BER (dashed lines) versus  $\bar{E}_b/\mathcal{N}_0$  for four different sets of 100 channels using Method I. UWB CM1 channel,  $R_c = 1/2$ , 16-QAM. No interference ( $\mathbf{J} = \mathbf{0}_{N \times 1}$ ).



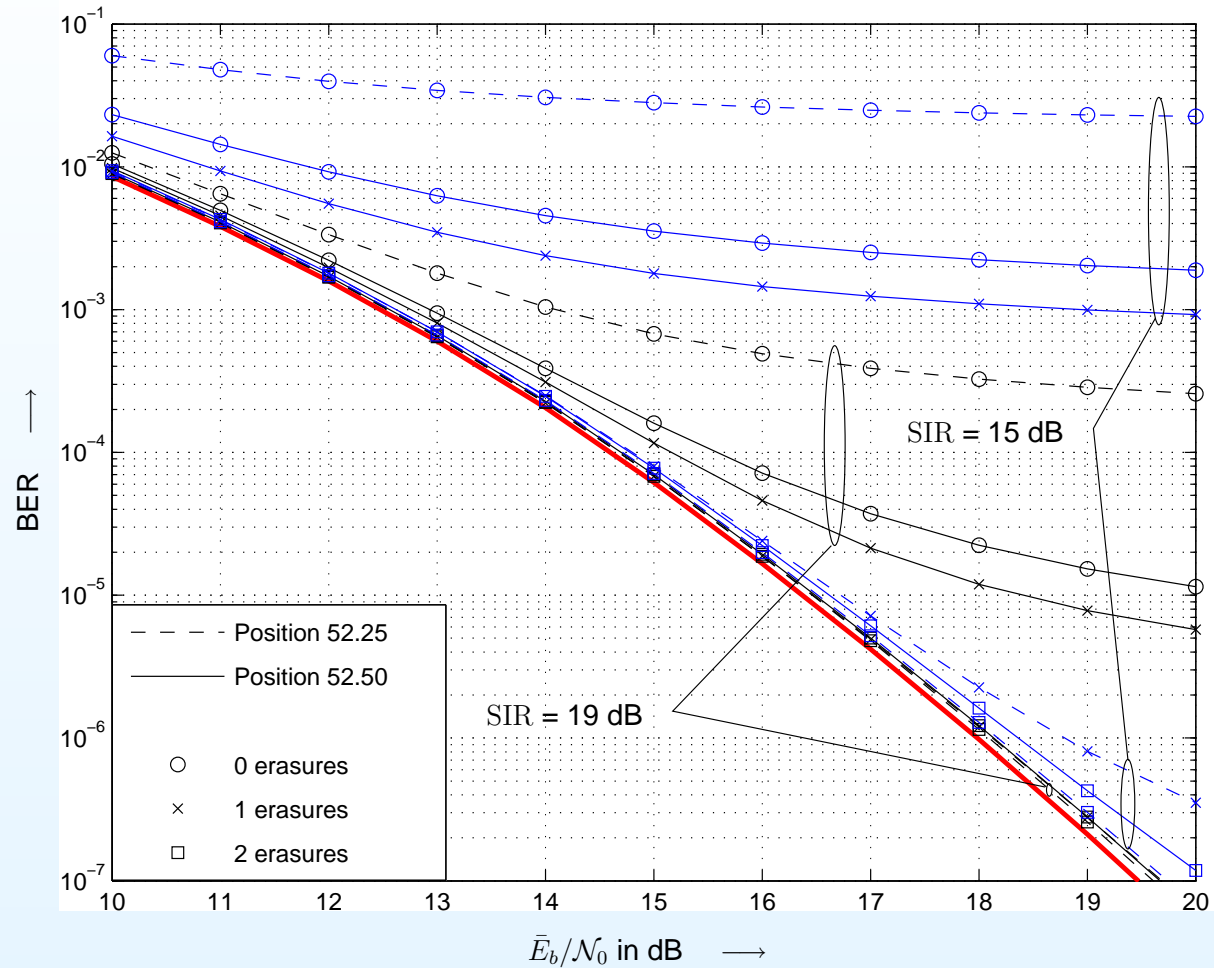
# Results: Average BER, Rayleigh Interference



Average BER versus  $\bar{E}_b/\mathcal{N}_0$  for various SIR with one Rayleigh-faded interferer, obtained using Method II. Interferer positions 52.0 (solid lines) and 52.5 (dashed lines). UWB CM1 channel,  $R_c = 1/2$ , 4-QAM.



# Results: Erasures



Average BER versus  $\bar{E}_b/\mathcal{N}_0$  for  $\{0, 1, 2\}$  subcarrier erasures. One non-faded interferer, positions 52.25 (dashed lines) and 52.5 (dashed lines) and SIR =  $\{15, 19\}$  dB, obtained using Method II. UWB CM1 channel,  $R_c = 1/2$ , 4-QAM.



## Conclusions

- Multiband OFDM: an excellent technology for Wireless PANs
- We have analysis methods to estimate
  - Outage BER (per-realization)
  - Average BER (per-realization, and direct)
  - Effect of interference (both methods)
- These methods apply to any multi-carrier system in quasi-static channel
- Tone interference may have a strong impact on MB-OFDM
- We can use erasure decoding to improve performance

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