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# Time-Domain Hybrid Pulse Amplitude Modulation

Power-Ratio and Ideal B2B Performance

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- BER Estimation for Square and Rectangular QAM;
  - Mathematical Formulation;
  - Numerical Results.

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  - Numerical Results.
  
- Hybrid PAM;
  - Mathematical Formulation;
  - BER Estimation;
  - Power-Ratio Management;
  - Numerical Results.

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- Hybrid PAM;
  - Mathematical Formulation;
  - BER Estimation;
  - Power-Ratio Management;
  - Numerical Results.
  
- Appendices
  - A - Mean Constellation Power;
  - B - Analytical Attempt to Determine the Optimum SNR for Minimum BER Operation;

# BER Estimation for Square and Rectangular QAM

# BER Estimation from SNR: Square QAM

- Most commonly, the BER of square QAM can be estimated from the SNR, by applying the following **closed-form expression**:

$$\Psi \left[ \text{SNR}, M \right] \simeq \frac{2}{\log_2(M)} \left( 1 - \frac{1}{\sqrt{M}} \right) \text{erfc} \left( \sqrt{\frac{3 \text{SNR}}{2(M-1)}} \right), \quad (1)$$

where

- $M$  - number of symbols in the constellation;
- SNR - SNR per symbol;
- $\text{erfc}(x) = \frac{2}{\sqrt{\pi}} \int_x^\infty e^{-t^2} dt$  is the complementary error function.

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- However, equation 1 neglects the bit errors originated at inter-symbol distances larger than the minimum Euclidean distance;
    - It therefore tends to **underestimate BER** in the low SNR (high BER) regime;

# BER Estimation from SNR: Square QAM

- An **exact formulation** for **BER estimation in an AWGN** channel has been derived in (1), as a **weighting average of all possible bit error probabilities** between all symbols in the QAM constellation:

$$\Psi \left[ \text{SNR}, M \right] = \frac{2}{\log_2(M)} \sum_{k=1}^{\log_2(\sqrt{M})} \Phi(k), \quad (2)$$



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where  $\Phi(k)$  is the  $k$ -th component of the overall BER, given by,

$$\Phi(k) = \frac{1}{\sqrt{M}} \sum_{i=0}^{(1-2^{-k})\sqrt{M}-1} \left\{ (-1)^{\lfloor \frac{i \cdot 2^{k-1}}{\sqrt{M}} \rfloor} \left( 2^{k-1} - \left\lfloor \frac{i \cdot 2^{k-1}}{\sqrt{M}} + \frac{1}{2} \right\rfloor \right) \operatorname{erfc} \left( (2i+1) \sqrt{\frac{3 \text{SNR}}{2(M-1)}} \right) \right\}, \quad (3)$$

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- For **high SNR**, the **first term ( $i = 0$ ) is dominant** in equation (3). By keeping only the first term of each  $\Phi(k)$  component, we arrive at the **approximated expression** of equation (1).

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- Note that, following the formulation of (2), the **exact BER of QPSK can be calculated by a single term** (the approximated formulation of (2) is exact for QPSK);
- Then, the number of BER components increases with the QAM constellation size.
  - $k \in [1, 2]$  for 16QAM;
  - $k \in [1, 2, 3]$  for 64QAM;
  - $k \in [1, 2, 3]$  for 64QAM;
  - ...

# BER Estimation from SNR: Rectangular QAM

- The exact BER expression for square QAM can be **generalized for any kind of rectangular QAM** (1):

$$\Psi \left[ \text{SNR}, M_I, M_Q \right] = \frac{1}{\log_2(M_I \cdot M_Q)} \left( \sum_{k=1}^{\log_2(M_I)} \Phi_I(k) + \sum_{n=1}^{\log_2(M_Q)} \Phi_Q(n) \right), \quad (4)$$

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- Similarly to the case of square QAM, an **approximated formula for high SNR** can be obtained by keeping only the  $i = 0$  and  $j = 0$  terms:

$$\Psi \left[ \text{SNR}, M_I, M_Q \right] \simeq \frac{1}{\log_2(M_I \cdot M_Q)} \left( \frac{M_I - 1}{M_I} + \frac{M_Q - 1}{M_Q} \right) \text{erfc} \left( \sqrt{\frac{3 \text{SNR}}{M_I^2 + M_Q^2 - 2}} \right), \quad (7)$$

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- Also note that, if  $M_I = M_Q = \sqrt{M}$  (square QAM), then **expression (7) reduces to equation (1)**;



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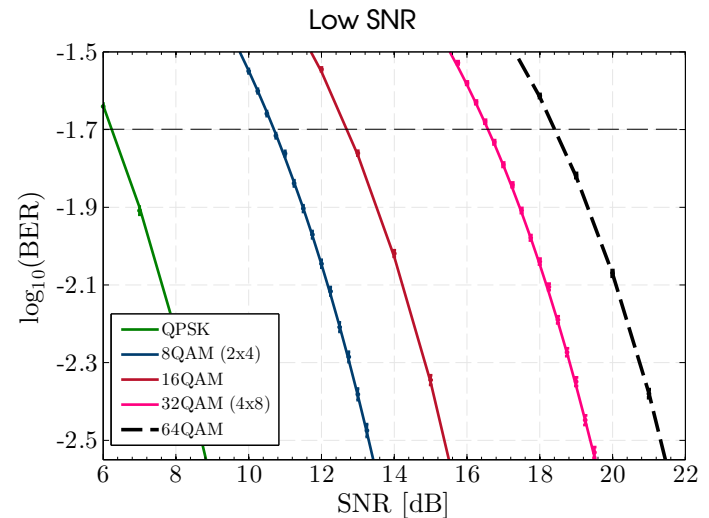
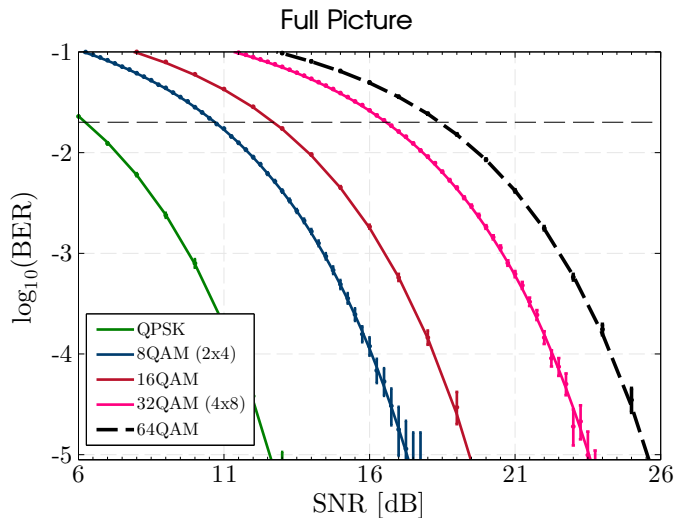
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- Also note that, if  $M_I = M_Q = \sqrt{M}$  (square QAM), then **expression (7)** reduces to **equation (1)**;
- These exact and approximated formulations allow for a **generalized treatment of any QAM** (square or rectangular) **or even PAM** ( $M_I \times 1$  or  $1 \times M_Q$ ) constellation.

# BER Estimation from SNR: Numerical Results

- We have ran **simulations in an AWGN channel**, and compared the **counted BER** with the **exact predictions** given by expression (4):



- There is a **very good agreement** both in high and low SNR regions;
- It is important to note, however, that the **rectangular 8QAM** and **32QAM** constellations are sub-optimal when compared with their cross-QAM implementation.

# Hybrid PAM - Mathematical Framework

# Hybrid PAM: Mathematical Formulation

- Let us assume a **general hybrid PAM frame**, composed of  $N_{\text{sym}}$  symbols and  $N_{\text{PAM}} \leq N_{\text{sym}}$  modulation formats, which can be completely described by:

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  - a  $[2 \times N_{\text{PAM}}]$  matrix defining the (square or rectangular) **QAM constellations**:

$$\mathbf{M} = \begin{bmatrix} M_I(1) & M_I(2) & \dots & M_I(N_{\text{PAM}}) \\ M_Q(1) & M_Q(2) & \dots & M_Q(N_{\text{PAM}}) \end{bmatrix}. \quad (8)$$

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- a  $[2 \times N_{\text{PAM}}]$  matrix defining the **power ratio** between formats (and between I and Q):

$$\mathbf{P} = \begin{bmatrix} P_I(1) & P_I(2) & \dots & P_I(N_{\text{PAM}}) \\ P_Q(1) & P_Q(2) & \dots & P_Q(N_{\text{PAM}}) \end{bmatrix}, \quad (9)$$

where  $P_I(n)/P_Q(n)$  defines the power ratio between I and Q in the  $n$ -th format and  $P_{I/Q}(n)/P_{I/Q}(k)$  defines the power ratio between the I/Q components of the  $n$ -th and  $k$ -th format in the frame;

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- a  $[1 \times N_{\text{PAM}}]$  vector defining the **time ratio** of each format in the frame:

$$\mathbf{F} = [F(1) \quad F(2) \quad \dots \quad F(N_{\text{PAM}})], \quad (10)$$

where  $F(n)$  defines the **time ratio occupied by the  $n$ -th format** in the frame. Note that  $\sum_{n=1}^{N_{\text{PAM}}} F(n) = 1$ .

# Hybrid PAM: Mathematical Formulation

- Associated with the **frame ratio matrix,  $\mathbf{F}$** , and the **number of symbols per frame,  $N_{\text{sym}}$** , we can define the following a **set of indices**:

$$[J(k)] = \left\{ j : N_{\text{sym}} \sum_{i=1}^{k-1} F(i) < j \leq N_{\text{sym}} \sum_{i=1}^k F(i), j \in \mathbb{N} \right\}, \quad (11)$$

where  $k$  is the modulation format index in the frame and  $[J(k)]$  are the indices of the corresponding symbols in the hybrid PAM frame.



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- This gives gives rise to the following  $[1 \times N_{\text{sym}}]$  **symbol index vector**:

$$\mathbf{J} = [[J(1)] \quad [J(2)] \quad \dots \quad [J(N_{\text{PAM}})]] . \quad (12)$$

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- In addition, note that the sets of possible discrete values for the **electrical field in the I and Q components** (normalized to unit power) are given by:

$$[I(M_I)] = \left\{ \frac{2k - \text{sgn}(k)}{\sqrt{\frac{2M_I^2 - 2}{3}}} : -\frac{M_I}{2} \leq k \leq \frac{M_I}{2}, k \in \mathbb{Z} \setminus \{0\} \right\}, \quad (13)$$

$$[Q(M_Q)] = \left\{ \frac{2k - \text{sgn}(k)}{\sqrt{\frac{2M_Q^2 - 2}{3}}} : -\frac{M_Q}{2} \leq k \leq \frac{M_Q}{2}, k \in \mathbb{Z} \setminus \{0\} \right\}. \quad (14)$$

# Hybrid PAM: Mathematical Formulation

- Utilizing the previously definitions, we can **mathematically describe the  $n$ -th sample of the electrical field,  $E(n)$** , of an hybrid PAM signal:

$$E(n) = \mathbf{P}^{(1,[J(k)])} \cdot [I(M_I(k))]^{(m_I(k))} + i\mathbf{P}^{(2,[J(k)])} \cdot [Q(M_Q(k))]^{(m_Q(k))}, \quad (15)$$

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- $\mathbf{P}^{(j,k)}$  denotes the element of  $\mathbf{P}$  located on the  $j$ -th row and  $k$ -th column;

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# Hybrid PAM - BER Estimation

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$$\Psi_{\text{TDHMF}} = \frac{\sum_{n=1}^{N_{\text{PAM}}} \log_2 (M_{\text{I}}(n) \cdot M_{\text{Q}}(n)) F(n) \cdot \Psi \left[ \text{SNR}(n), M_{\text{I}}(n), M_{\text{Q}}(n) \right]}{\sum_{n=1}^{N_{\text{PAM}}} \log_2 (M_{\text{I}}(n) \cdot M_{\text{Q}}(n)) F(n)} \quad (17)$$

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  - **same BER performance** for all formats at a given target BER;
  - **minimum overall BER** for a given average SNR.

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- In order to guarantee **constant power between modulation formats** in the hybrid PAM frame (and still guarantee the **same Euclidean distance between I and Q components** in each format), the power ratio matrix,  $\mathbf{P}$ , should be defined as follows:

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- Note that the **ratio between the numerators** of each  $\mathbf{P}$  row guarantees the **same Euclidean distance** between I and Q;
- In turn, the **ratio between the denominators** of each  $\mathbf{P}$  column guarantees **constant power** between the hybrid PAM symbols.



# Hybrid PAM: Power-Ratio - Same Euclidean Distance

- Since in the  $\mathbf{P}$  definition of expression (21) all quadratures are **normalized to unit power**, then it is a sufficient condition to guarantee the **same Euclidean distance** between all modulation formats in the frame.
- The power-ratio matrix,  $\mathbf{P}$ , is therefore rewritten here:

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# Hybrid PAM: Power-Ratio - Same BER

- In order to guarantee the **same BER performance at a given target BER**, one must adjust the power-ratio between modulation formats so that the SNR perceived by each one of them is given by:

$$\text{SNR}(n) = \Psi^{-1} \left[ \text{BER}, M_I(n), M_Q(n) \right] \quad (24)$$

where  $\Psi^{-1}[\cdot]$  is the **inverse of approximate BER estimation expression (7)**. Note that obtaining an inverse of the exact BER expression (4) would require to invert a series of  $\text{erfc}(\cdot)$  functions, which may be unfeasible in practice.

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- Note that, **relatively to the constant-power condition**, the same BER operation only requires to **multiply each P by the respective required SNR**.

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- To achieve the **theoretical minimum BER** requires to **minimize the BER estimation expression** given by (17), which is composed by  $N_{\text{PAM}}$   $\text{erfc}(\cdot)$  terms;

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- Once the **optimum SNR is found for each modulation format**, the power ratio matrix is defined similarly to (25):

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# Hybrid PAM: Power-Ratio - Minimum BER

- To achieve the **theoretical minimum BER** requires to **minimize the BER estimation expression** given by (17), which is composed by  $N_{\text{PAM}}$   $\text{erfc}(\cdot)$  terms;
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- The essential **difference between the "same BER" and "minimum BER"** strategies lies on the **determination of the SNR** associated to each modulation format:
  - **same BER**: the required SNR per format is **independently determined**;
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  - **same BER**: the required SNR per format is **independently determined**;
  - **minimum BER**: the optimum SNR per format requires a **joint optimization process**.
- In general, to minimize the BER of a frame composed of  $N_{\text{PAM}}$  format requires a  $N_{\text{PAM}} - 1$ -dimensional optimization process;
- This means that, in practice, if the frame is composed of more than two formats, then the **optimization process becomes quite complex**.

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$$\text{SNR}(2) = \frac{\overline{\text{SNR}} - F(1)\text{SNR}(1)}{F(2)} \quad (27)$$

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- The estimated BERs are then calculated via the **exact expression (4)**;
- The **optimum SNRs** are then obtained as:

$$\text{SNR}_{\text{opt}}(1) = \left\{ \text{SNR}(1) : \Psi_{\text{TDHMF}} \left[ \text{SNR}(1), \text{SNR}(2) \right] = \min \left\{ \Psi_{\text{TDHMF}} \left[ \text{SNR}(1), \text{SNR}(2) \right] \right\} \right\} \quad (28)$$

$$\text{SNR}_{\text{opt}}(2) = \frac{\overline{\text{SNR}} - F(1)\text{SNR}_{\text{opt}}(1)}{F(2)} \quad (29)$$

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- Alternatively, it is also possible to **analytically find the minimum BER** by evaluating the **derivative of the approximate BER estimation formula** of expression (7):

$$\frac{\partial \Psi_{\text{TDHMF}} \left[ \text{SNR}(1), \frac{\overline{\text{SNR}} - F(1)\text{SNR}(1)}{F(2)} \right]}{\partial \text{SNR}1} = 0, \quad (30)$$

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where

- $x = \text{SNR}(1)$ ;
- $\alpha = e^{2B_2} B_2$ ;
- $\beta = e^{2B_2} C_2$ ;
- $\kappa = -2(C_1 + C_2)$ ;
- $\gamma = \frac{A_2^3 C_2^2}{A_1^2 C_1}$ ;
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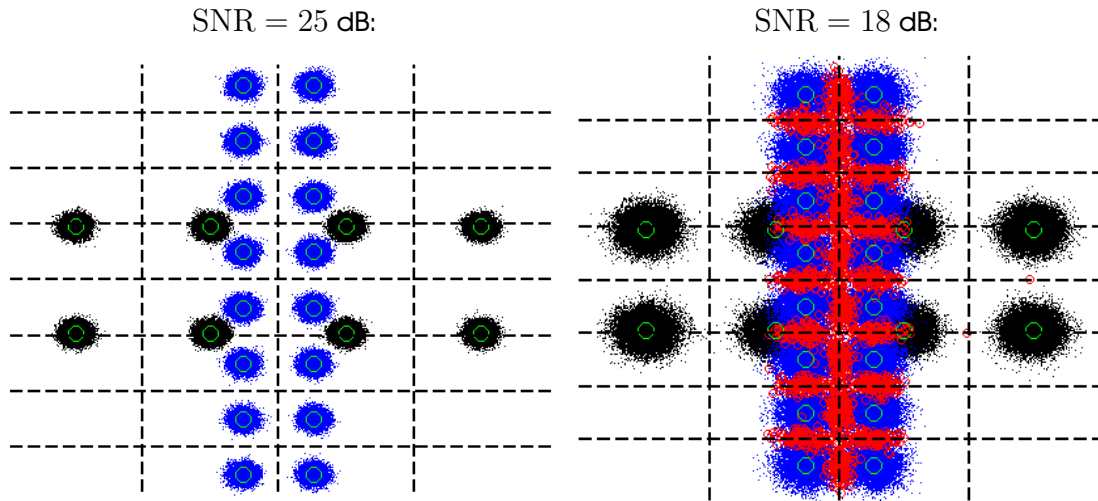
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- Unfortunately, my math skills do not allow me to find a closed-form solution for equation (31)... A numerical optimization process is still needed to find  $x$ .

# Some Examples: Same Power

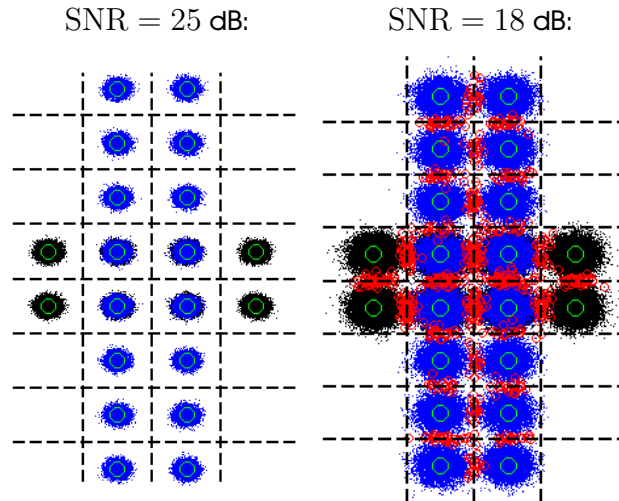
- Example of an exotic hybrid PAM constellation composed of  $4 \times 2\text{QAM}$  and  $2 \times 8\text{QAM}$ ;
- The power ratio between formats has been defined based on the **same power** criterion.



- Clearly, one of the formats is being **strongly penalized** relatively to the other.

# Some Examples: Same Euclidean Distance

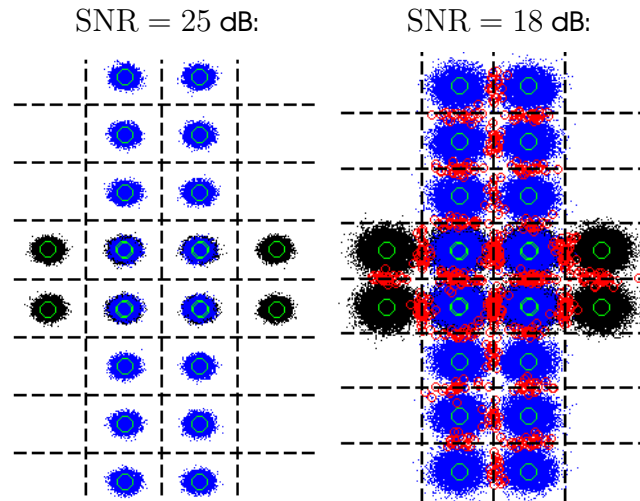
- Example of an exotic hybrid PAM constellation composed of  $4 \times 2\text{QAM}$  and  $2 \times 8\text{QAM}$ ;
- The power ratio between formats has been defined based on the **same Euclidean distance** criterion.



- **All symbols** have the same probability of error. However, this may be sub-optimal, since the two formats carry different number of bits per symbol.

# Some Examples: Same BER

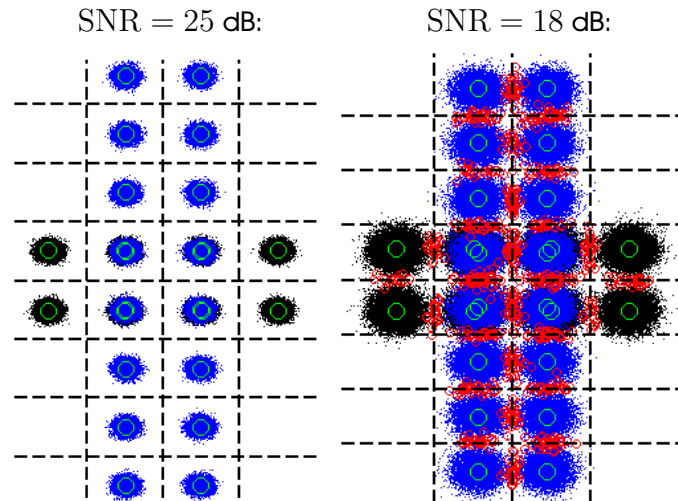
- Example of an exotic hybrid PAM constellation composed of  $4 \times 2\text{QAM}$  and  $2 \times 8\text{QAM}$ ;
- The power ratio between formats has been defined based on the **same BER** criterion.



- The probability of error is now the same for both formats. It is still sub-optimum, since the two formats carry **different number of bits per symbol**.

# Some Examples: Minimum BER

- Example of an exotic hybrid PAM constellation composed of  $4 \times 2\text{QAM}$  and  $2 \times 8\text{QAM}$ ;
- The power ratio between formats has been defined based on the **minimum BER** criterion.

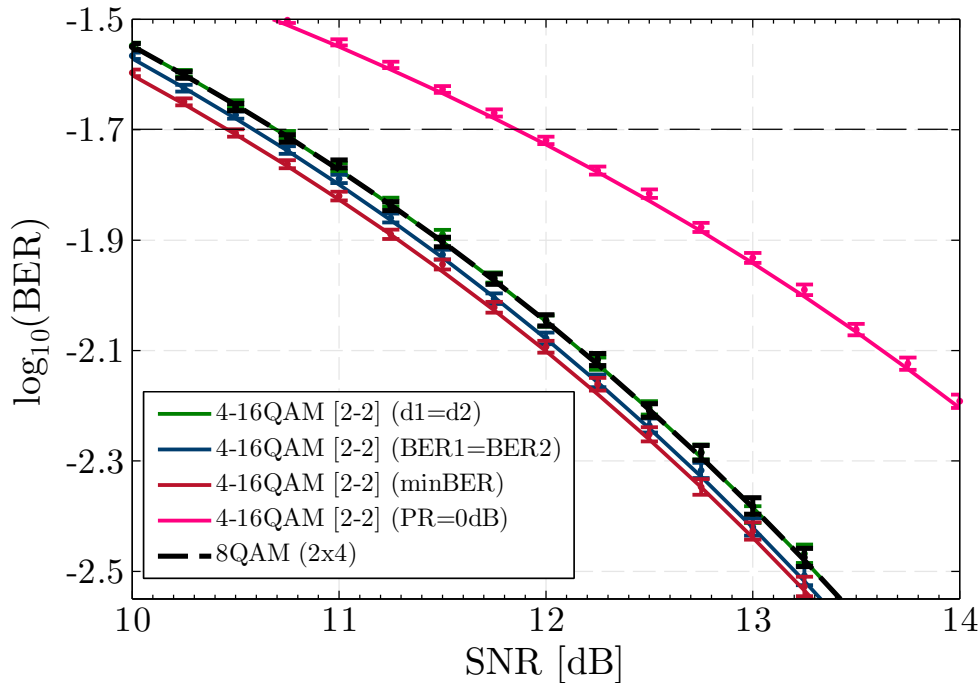


- The BER of the ensemble hybrid PAM frame is minimized - **optimum solution**.
- Note that the power-ratio between formats actually depends on the SNR.

# Hybrid PAM - Numerical Results

# Hybrid 4/16QAM: Numerical Results

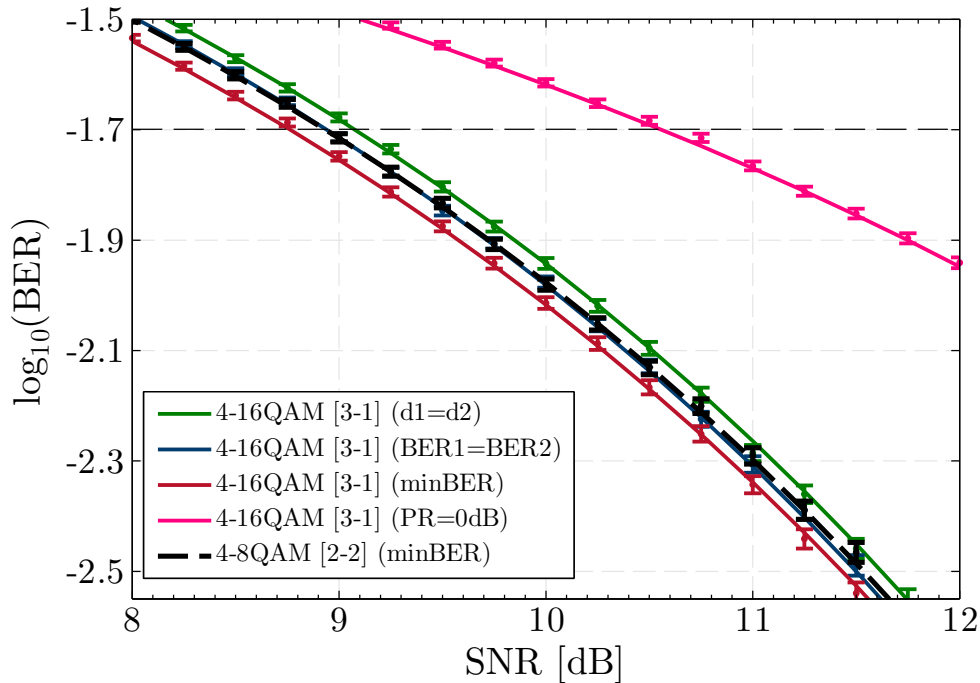
- Performance of Hybrid 4-16QAM with 50%-50% power ratio (3 bits per symbol):



- Good agreement between numerical results and analytical BER estimation;
- Good agreement with the OFC 2014 results (except for the 8QAM, which is rectangular instead of star QAM).

# Hybrid 4/16QAM: Numerical Results

- Performance of Hybrid 4-16QAM with 75%-25% power ratio (2.5 bits per symbol):

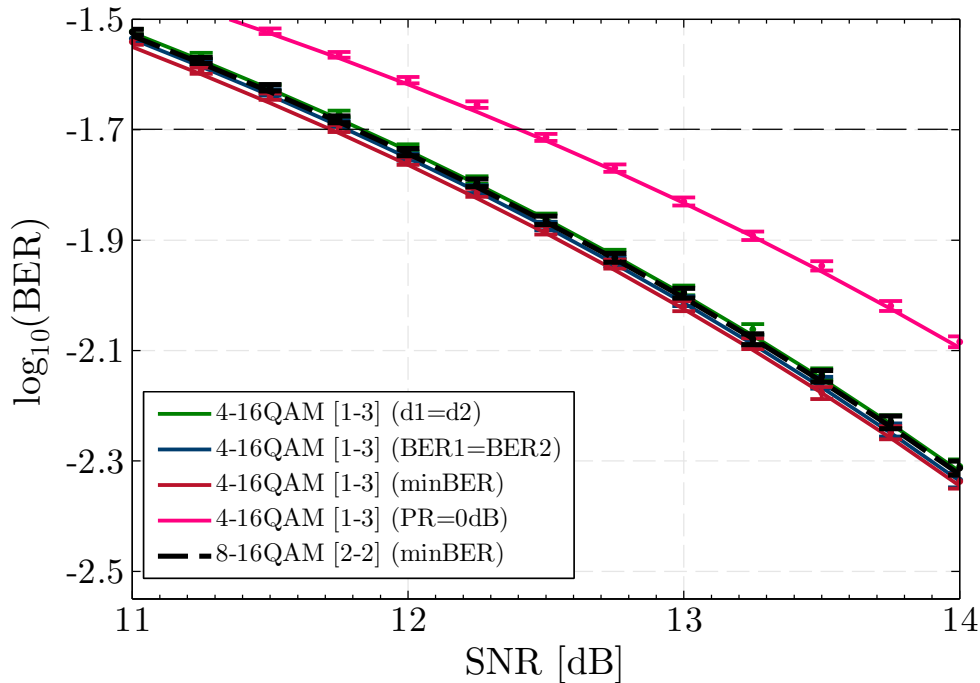


- Hybrid 4-16QAM (75%-25%) is a slightly better solution than Hybrid 4-8QAM (50%-50%);
- Most probably due to the sub-optimum performance of rectangular 8QAM.



# Hybrid 4/16QAM: Numerical Results

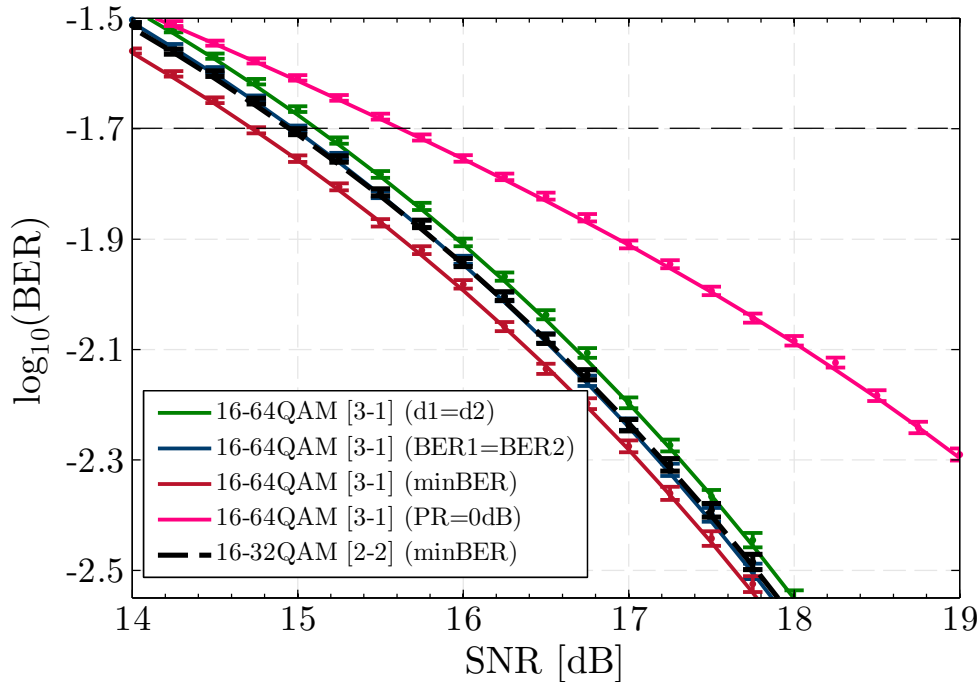
- Performance of Hybrid 4-16QAM with 25%-75% power ratio (3.5 bits per symbol):



- Hybrid 4-16QAM (25%-75%) is a slightly better solution than Hybrid 8-16QAM (50%-50%);
- Most probably due to the sub-optimum performance of rectangular 8QAM.

# Hybrid 16/64QAM: Numerical Results

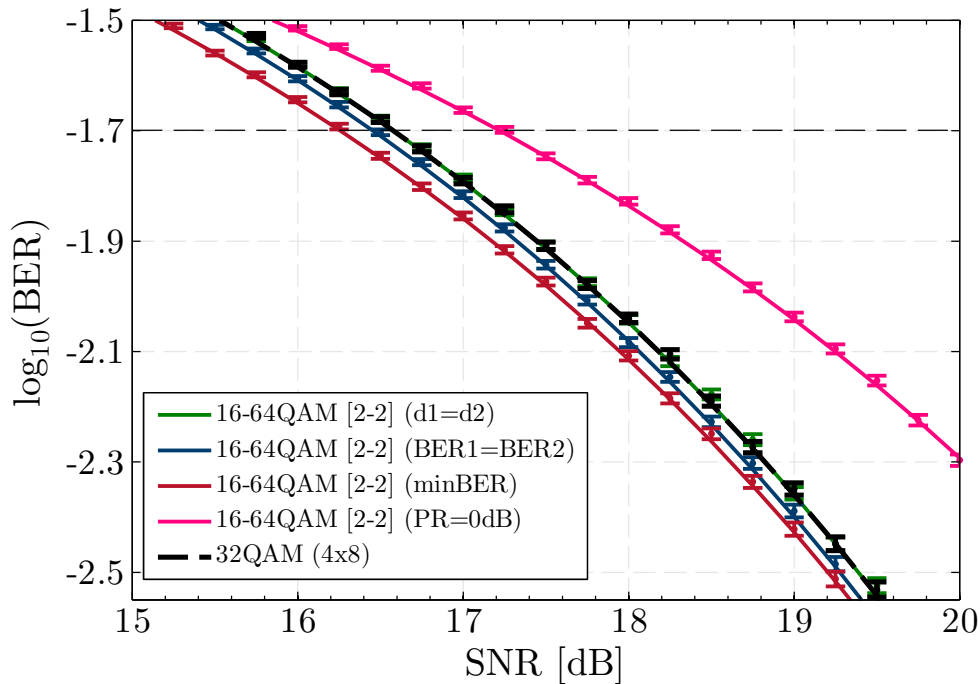
- Performance of Hybrid 16-64QAM with **75%-25%** power ratio (**4.5 bits per symbol**):



- Hybrid 16-64QAM (75%-25%) is a slightly better solution than Hybrid 16-32QAM (50%-50%);
- Most probably due to the sub-optimum performance of rectangular 32QAM.

# Hybrid 16/64QAM: Numerical Results

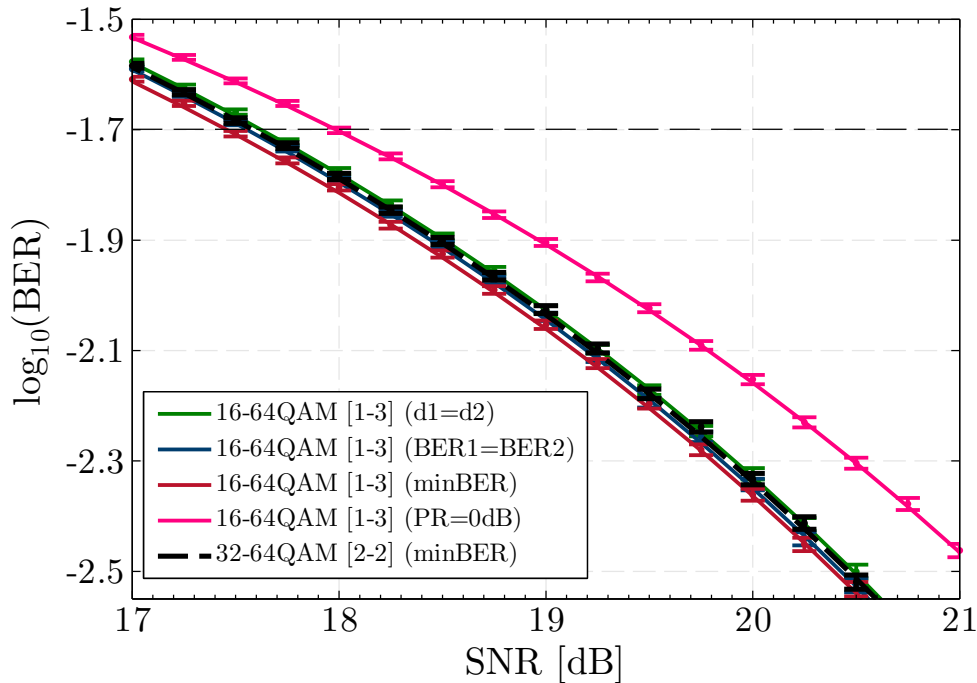
- Performance of Hybrid 16-64QAM with **50%-50%** power ratio (**5 bits per symbol**):



- Hybrid 16-64QAM (50%-50%) is a slightly better solution than 32QAM;
- Most probably due to the sub-optimum performance of rectangular 32QAM.

# Hybrid 16/64QAM: Numerical Results

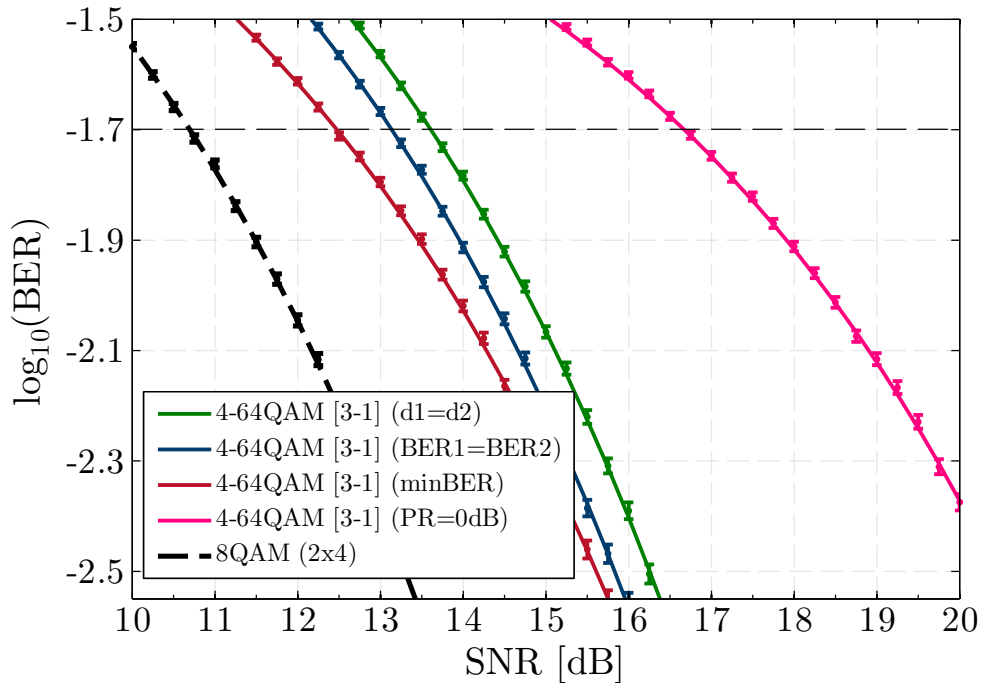
- Performance of Hybrid 16-64QAM with **25%-75%** power ratio (**5.5 bits per symbol**):



- Hybrid 16-64QAM (25%-75%) is a slightly better solution than Hybrid 32-64QAM (50%-50%);
- Most probably due to the sub-optimum performance of rectangular 32QAM.

# Hybrid 4/64QAM: Numerical Results

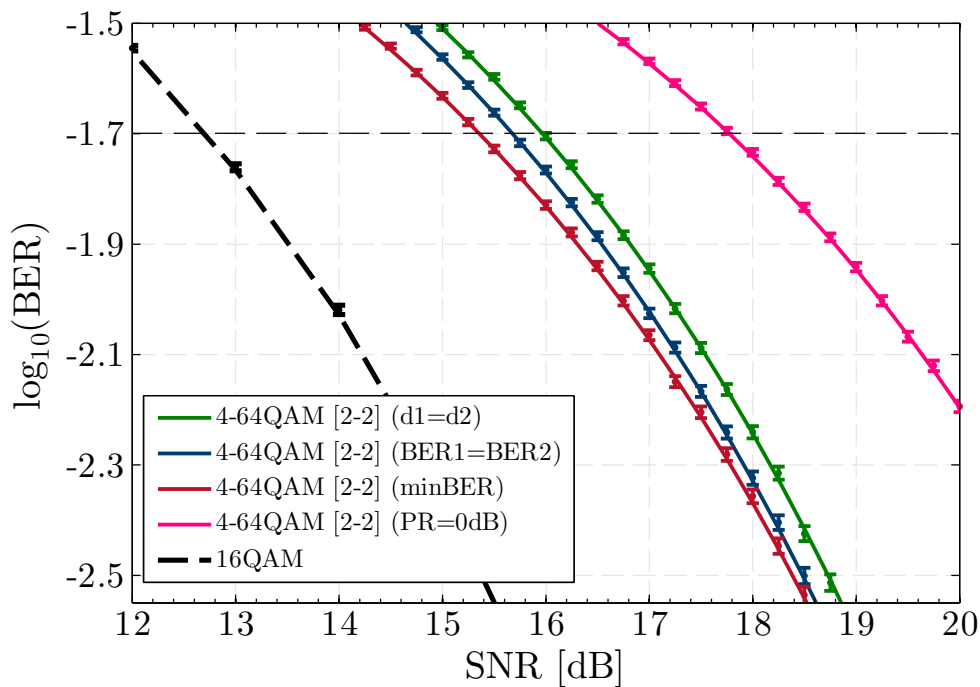
- Performance of Hybrid 4-64QAM with 75%-25% power ratio (3 bits per symbol):



- Hybrid 4-64QAM (75%-25%) is a much worse solution than 8QAM ( $\sim 2.5$  dB penalty);
- It seems that hybridizing formats with very different constellation size is not advantageous.

# Hybrid 4/64QAM: Numerical Results

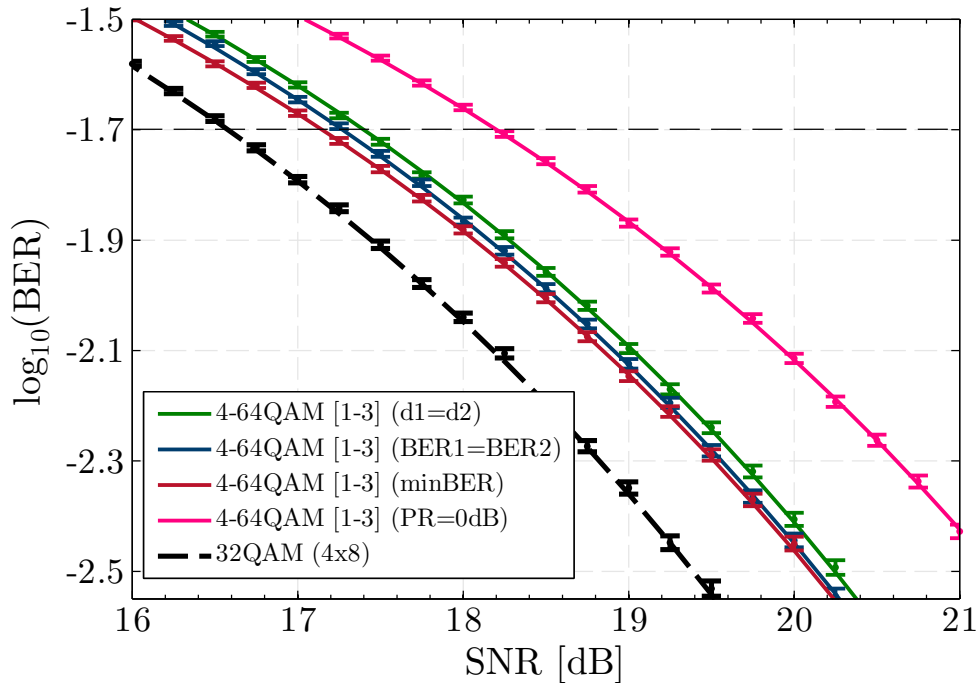
- Performance of Hybrid 4-64QAM with 50%-50% power ratio (4 bits per symbol):



- Hybrid 4-64QAM (50%-50%) is a much worse solution than 16QAM ( $\sim 2.5$  dB penalty);
- It seems that hybridizing formats with very different constellation size is not advantageous.

# Hybrid 4/64QAM: Numerical Results

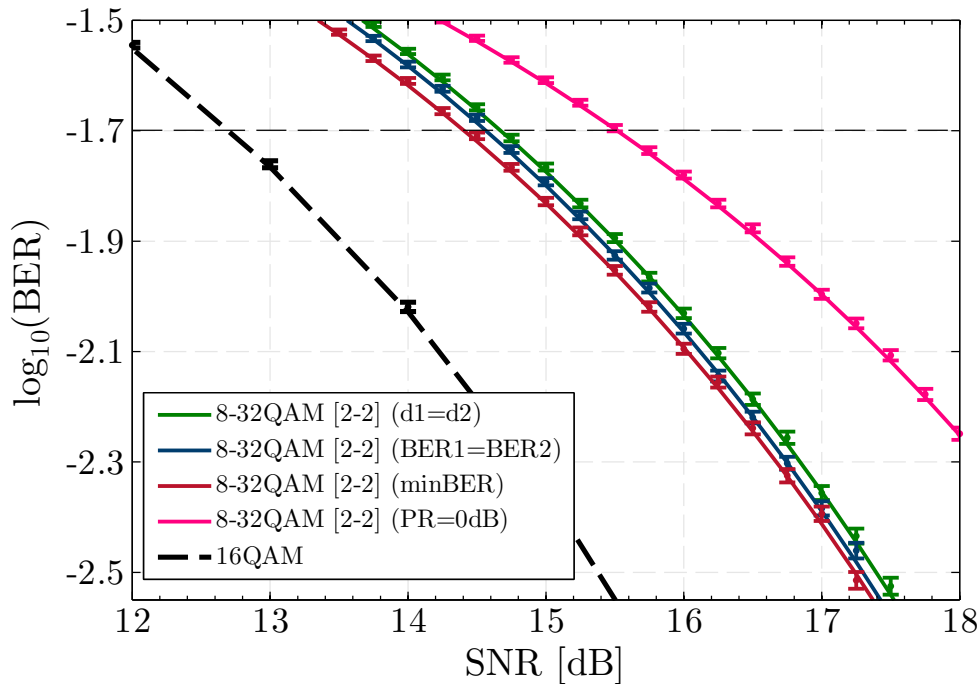
- Performance of Hybrid 4-64QAM with 25%-75% power ratio (5 bits per symbol):



- Hybrid 4-64QAM (25%-75%) yields a penalty of  $\sim 0.6$  dB over 32QAM;
- Surprisingly, the penalty is significantly lower with this configuration.

# Hybrid 8/32QAM: Numerical Results

- Performance of Hybrid 8-32QAM with 50%-50% power ratio (4 bits per symbol):



- Hybrid 8-32QAM (50%-50%) yields a penalty of  $\sim 1.5$  dB over 16QAM;
- Maybe in this case the penalty is mostly due the sub-optimum performance of 8QAM and 32QAM.



# Appendix A - Average Constellation Power

# Average Constellation Power of Rectangular QAM

- Any rectangular QAM constellation (if centered in the origin) is fully symmetric over the 4 quadrants of the complex plane;
- To analyze the average power of a general rectangular QAM constellation, we may double-fold it over one of its quadrants:

$$\begin{aligned}\bar{P} &= \frac{4}{M_I \cdot M_Q} \sum_{Q=1}^{M_Q/2} \sum_{I=1}^{M_I/2} [(2I-1)^2 + (2Q-1)^2] \\ &= \frac{4}{M_I \cdot M_Q} \sum_{Q=1}^{M_Q/2} \left[ \frac{M_I}{2} (2Q-1)^2 + \sum_{I=1}^{M_I/2} (2I-1)^2 \right] \\ &= \frac{4}{M_I \cdot M_Q} \sum_{Q=1}^{M_Q/2} \left[ \frac{M_I}{2} (2Q-1)^2 + \frac{M_I^3 - M_I}{6} \right] \\ &= \frac{4}{M_I \cdot M_Q} \frac{M_I}{2} \sum_{Q=1}^{M_Q/2} (2Q-1)^2 + \frac{M_Q}{2} \frac{M_I^3 - M_I}{6} \\ &= \frac{4}{M_I \cdot M_Q} \frac{M_I(M_Q^3 - M_Q) + M_Q(M_I^3 - M_I)}{12} = \frac{M_I^2 + M_Q^2 - 2}{3}\end{aligned}\tag{32}$$

# Average Constellation Power of Rectangular QAM

- If the I and Q components are divided by general factors  $C_I$  and  $C_Q$ , respectively, expression (32) can be generalized as:

$$\bar{P} = \frac{M_I^2 - 1}{3C_I^2} + \frac{M_Q^2 - 1}{3C_Q^2} \quad (33)$$

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- Now, let us consider that the I and Q components are normalized to unit power (in the complex plane) and then scaled according to the power ratio defined in the  $\mathbf{P}$  matrix. The corresponding mean constellation power is given by:

$$\begin{aligned} \bar{P} &= \frac{M_I^2 - 1}{3 \left( \frac{\sqrt{\frac{2M_I^2 - 2}{3}}}{\sqrt{P_I}} \right)^2} + \frac{M_Q^2 - 1}{3 \left( \frac{\sqrt{\frac{2M_Q^2 - 2}{3}}}{\sqrt{P_Q}} \right)^2} \\ &= P_I \frac{M_I^2 - 1}{2M_I^2 - 2} + P_Q \frac{M_Q^2 - 1}{2M_Q^2 - 2} \\ &= \frac{P_I + P_Q}{2} \end{aligned} \quad (34)$$

# **Appendix B - Analytical Attempt to Determine the Optimum SNR for Minimum BER Operation**

# Analytical Determination of Optimum SNR

- Determining the optimum SNR for the first modulation format,  $\text{SNR}(1)$ , in a frame composed of 2 modulation formats ( $N_{\text{PAM}} = 2$ ) involves solving the following differential equation:

$$\frac{\partial \Psi_{\text{TDHMF}} \left[ \text{SNR}(1), \frac{\overline{\text{SNR}} - F(1)\text{SNR}(1)}{F(2)} \right]}{\partial \text{SNR}(1)} = 0$$
$$\Leftrightarrow F(1) \log_2 (M_I(1)M_Q(1)) \frac{\partial \Psi [\text{SNR}(1)]}{\partial \text{SNR}(1)} = -F(2) \log_2 (M_I(2)M_Q(2)) \frac{\partial \Psi \left[ \frac{\overline{\text{SNR}} - F(1)\text{SNR}(1)}{F(2)} \right]}{\partial \text{SNR}(1)}, \quad (35)$$

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- The derivative of the  $\text{erfc}(\cdot)$  function is defined and the following derivative pairs can be used to solve (35):

$$\frac{\partial}{\partial x} \text{erfc}(\sqrt{c_1 x}) = -\frac{c_1}{\sqrt{c_1 \pi x}} e^{-c_1 x}, \quad (36)$$

$$\frac{\partial}{\partial x} \text{erfc}(\sqrt{c_2 - c_1 x}) = \frac{c_1}{\sqrt{\pi(c_2 - c_1 x)}} e^{-c_2 + c_1 x}, \quad (37)$$

# Analytical Determination of Optimum SNR

- To simplify the notation, let us define the following constants:

- $A_1 = F(1) \left( \frac{M_I(1) - 1}{M_I(1)} + \frac{M_Q(1) - 1}{M_Q(1)} \right);$

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- $A_2 = F(2) \left( \frac{M_I(2) - 1}{M_I(2)} + \frac{M_Q(2) - 1}{M_Q(2)} \right);$

- $B_2 = 3 \frac{\text{SNR}}{F(2) (M_I(2)^2 + M_Q(2)^2 - 2)};$

- $C_1 = \frac{3}{(M_I(1)^2 + M_Q(1)^2 - 2)};$

- $C_2 = \frac{3F(1)}{F(2) (M_I(2)^2 + M_Q(2)^2 - 2)}.$

- Equation (35) then reduces to

$$- \frac{A_1 C_1}{\sqrt{C_1 \pi \text{SNR}(1)}} e^{-C_1 \text{SNR}(1)} = \frac{A_2 C_2}{\sqrt{\pi (B_2 - C_2 \text{SNR}(1))}} e^{-B_2 + C_2 \text{SNR}(1)} \quad (38)$$

# Analytical Determination of Optimum SNR

- To simplify the notation, let us define the following constants:

- $A_1 = F(1) \left( \frac{M_I(1) - 1}{M_I(1)} + \frac{M_Q(1) - 1}{M_Q(1)} \right);$

- $A_2 = F(2) \left( \frac{M_I(2) - 1}{M_I(2)} + \frac{M_Q(2) - 1}{M_Q(2)} \right);$

- $B_2 = 3 \frac{\text{SNR}}{F(2) (M_I(2)^2 + M_Q(2)^2 - 2)};$

- $C_1 = \frac{3}{(M_I(1)^2 + M_Q(1)^2 - 2)};$

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- Since  $\text{SNR}(1)$  and all constants in (38) are supposed to be positive, we can apply the  $(\cdot)^2$  operation to both sides of the equation without loss of generality, yielding:

$$\left[ \frac{e^{2B_2} B_2}{\text{SNR}(1)} - e^{2B_2} C_2 \right] e^{-2(C_1 + C_2) \text{SNR}(1)} = \frac{A_2^2 C_2^2}{A_1^2 C_1} \quad (39)$$

# Analytical Determination of Optimum SNR

- Equation (39) is a transcendental equation of the form

$$\left[ \frac{\alpha}{x} - \beta \right] e^{\kappa x} = \gamma, \quad (40)$$

with

- $x = \text{SNR}(1)$ ;
- $\alpha = e^{2B_2} B_2$ ;
- $\beta = e^{2B_2} C_2$ ;
- $\kappa = -2(C_1 + C_2)$ ;
- $\gamma = \frac{A_2^2 C_2^2}{A_1^2 C_1}$ ;

which apparently does not lend itself to a closed-form solution...

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- Anyway, equation (39) can be solved numerically, in a similar way as described in (28), with the advantage of not requiring the computation of  $\text{erfc}(\cdot)$  functions.

# Thanks for your attention!

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