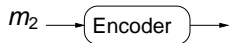
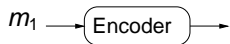


# The Degrees of Freedom of Compute-and-Forward

Urs Niesen  
Jointly with Phil Whiting

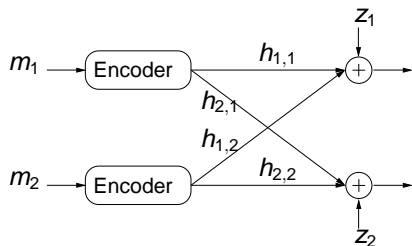
Bell Labs, Alcatel-Lucent

# Problem Setting



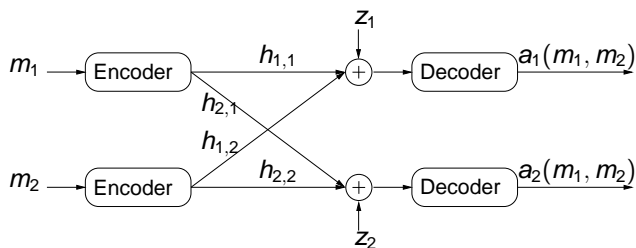
- $K$  transmitters, messages  $m_1, \dots, m_K$ , power constraint  $P$

# Problem Setting



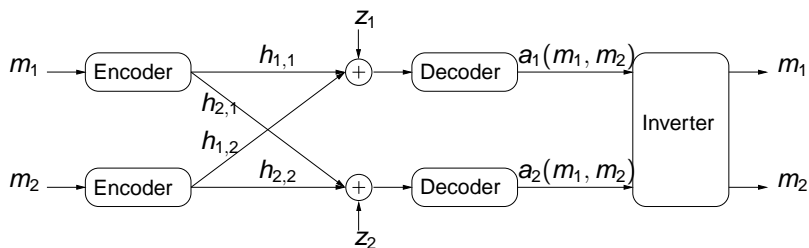
- $K$  transmitters, messages  $m_1, \dots, m_K$ , power constraint  $P$
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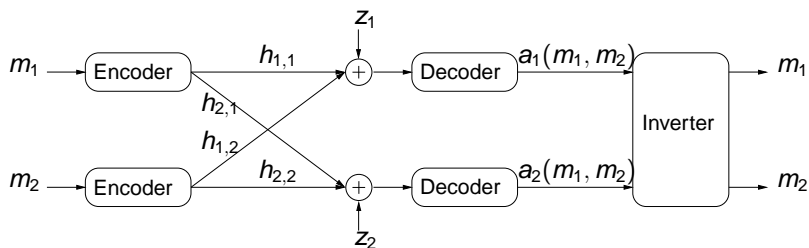
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# Problem Setting



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- **Invert** all computed functions  $a_1, \dots, a_K$  to recover the messages  $m_1, \dots, m_K$

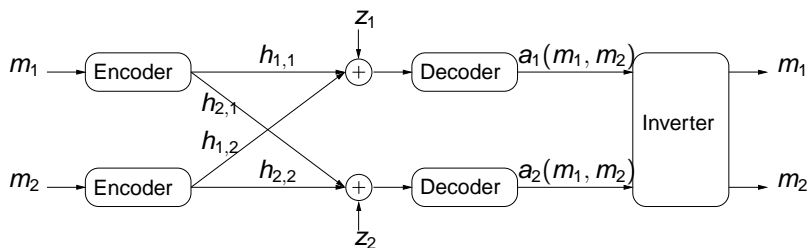
# Problem Setting



- Computation capacity  $C(P, \mathbf{H}, \{a_\ell\})$  for **fixed** function  $\{a_\ell\}$



# Problem Setting



- Computation capacity  $C(P, \mathbf{H}, \{a_\ell\})$  for **fixed** function  $\{a_\ell\}$
- $C(P, \mathbf{H}, \mathbf{A})$  for **linear** function  $\{a_\ell\}$
- Computation capacity

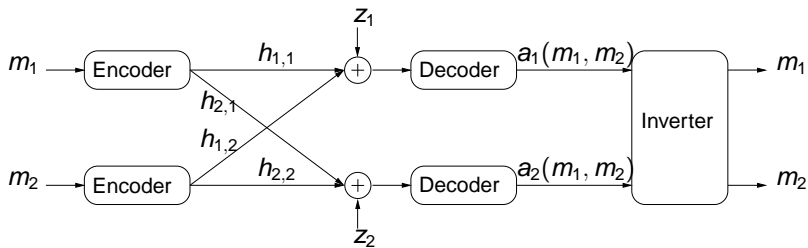
$$C(P, \mathbf{H}) \triangleq \max_{\{a_\ell\}} C(P, \mathbf{H}, \{a_\ell\})$$

with maximization over all invertible functions  $\{a_\ell\}$



# Computation Capacity

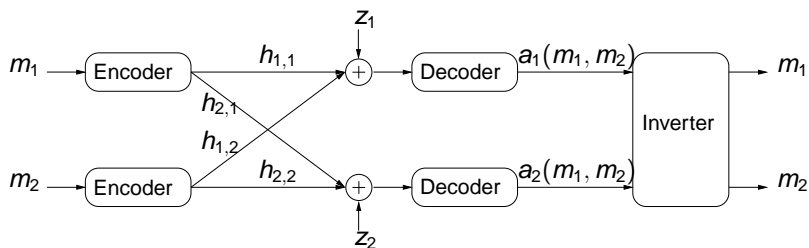
## Some Special Cases and Bounds



- Identity function  $\mathbf{A} = \mathbf{I} \Rightarrow C(P, \mathbf{H}, \mathbf{I})$  is the capacity of the  $K$ -user interference channel

# Computation Capacity

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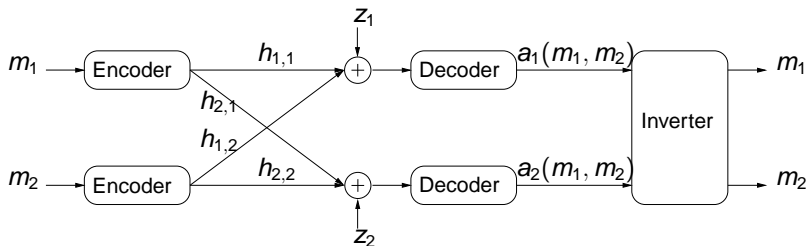


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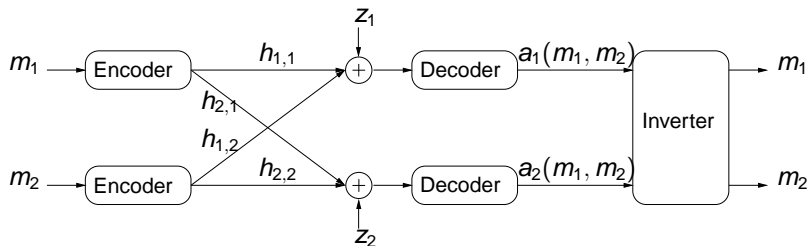


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$$C(P, \mathbf{H}) \geq C(P, \mathbf{H}, \mathbf{I}) = \frac{K}{4} \log(P) + o(\log(P))$$

# Computation Capacity

## Some Special Cases and Bounds

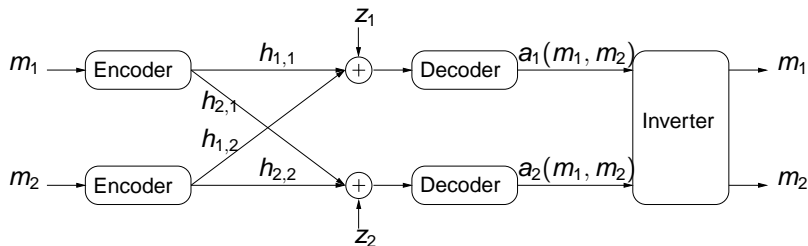


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# Computation Capacity

## Some Special Cases and Bounds



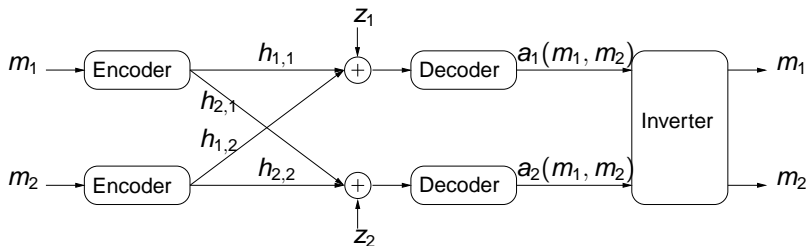
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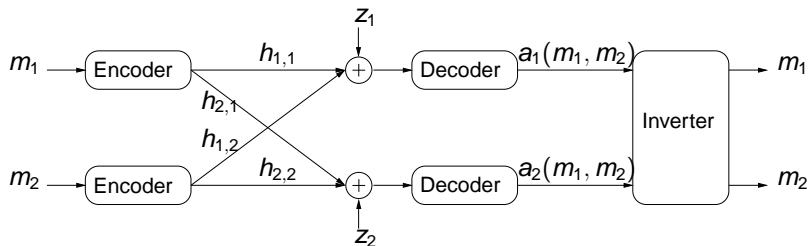
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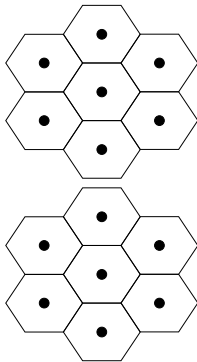
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# Computation Capacity

## Lattice Codes

Nazer and Gastpar (2009)



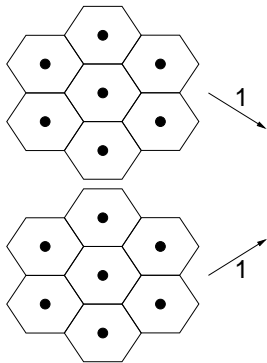
- Two transmitters, one receiver,  $\mathbf{h} = (1, 1)$



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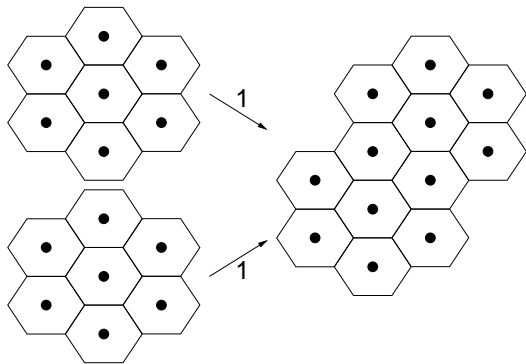


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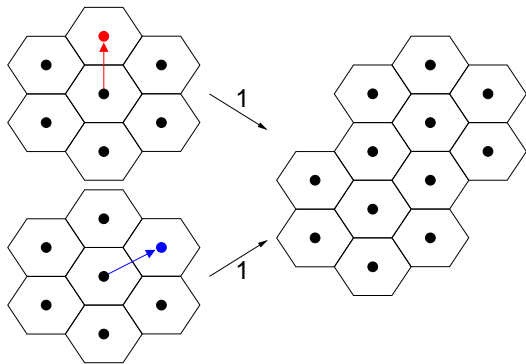


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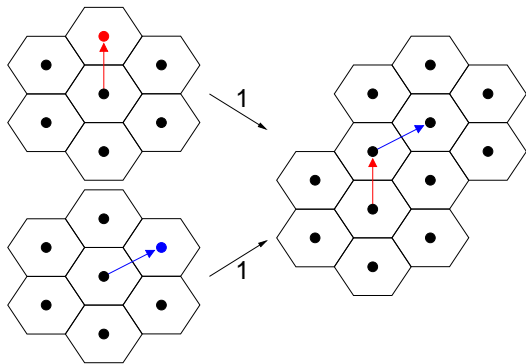


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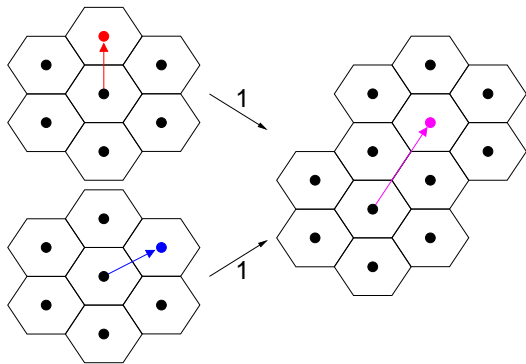


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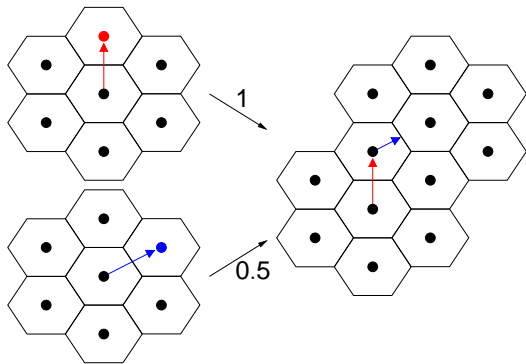


- Two transmitters, one receiver,  $\mathbf{h} = (1, 1)$
- Decode  $\mathbf{a} = (1, 1)$

# Computation Capacity

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Nazer and Gastpar (2009)

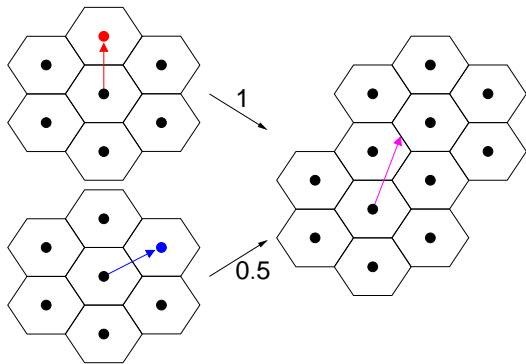


- Two transmitters, one receiver,  $\mathbf{h} = (1, 0.5)$

# Computation Capacity

## Lattice Codes

Nazer and Gastpar (2009)



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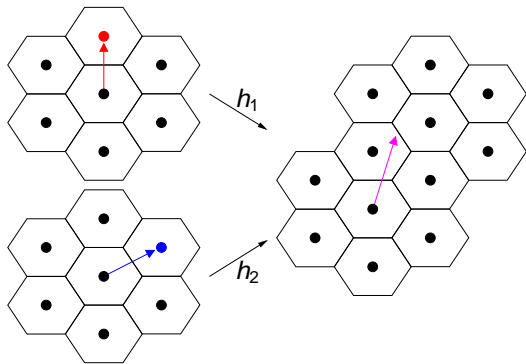
- Two transmitters, one receiver,  $\mathbf{h} = (1, 0.5)$
- Scale output by  $\beta = 2$ , decode  $\mathbf{a} = (2, 1)$



# Computation Capacity

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Nazer and Gastpar (2009)

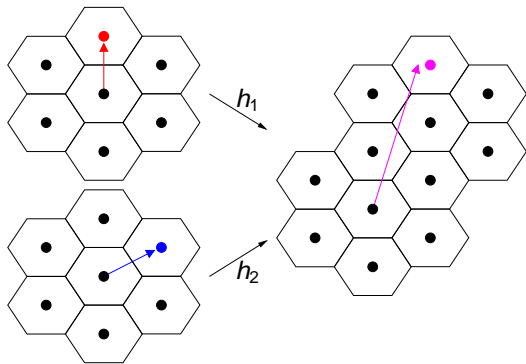


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Nazer and Gastpar (2009)

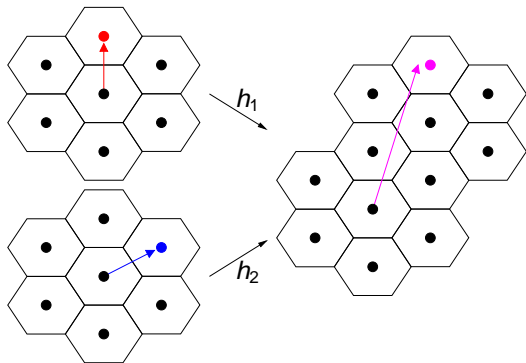


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# Computation Capacity

Lattice Codes

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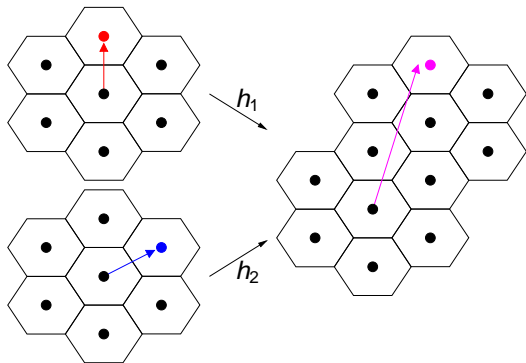
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# Computation Capacity

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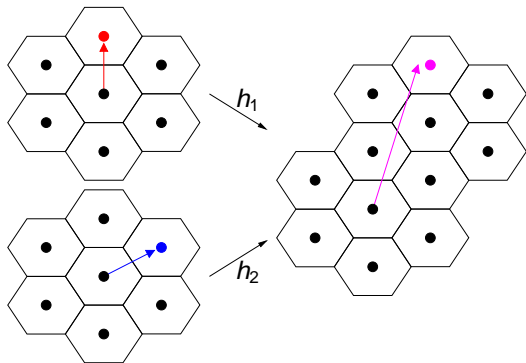
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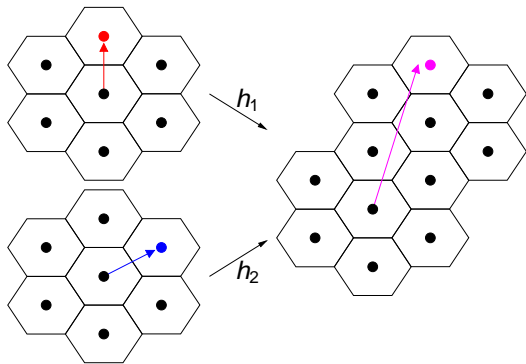
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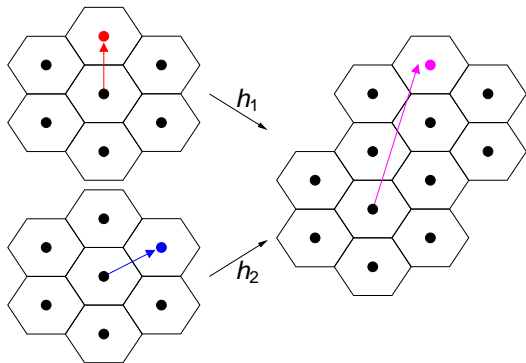


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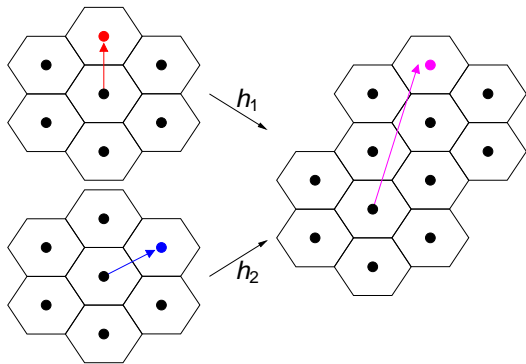
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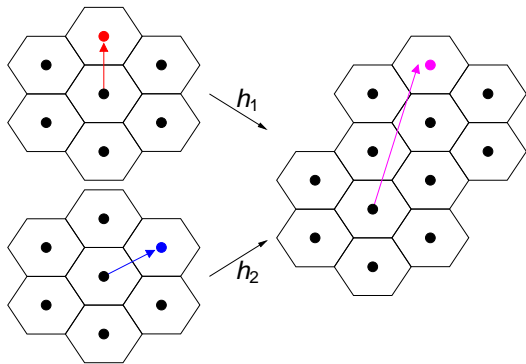
$$\max_{\mathbf{A}} C(P, \mathbf{H}, \mathbf{A}) \geq \max_{\mathbf{A}} R_L(P, \mathbf{H}, \mathbf{A})$$



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# Questions

- We already know that the computation capacity satisfies

$$K/2 \leq \lim_{P \rightarrow \infty} \frac{C(P, \mathbf{H})}{\frac{1}{2} \log(P)} \leq K$$

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$$\lim_{P \rightarrow \infty} \frac{C(P, \mathbf{H})}{\frac{1}{2} \log(P)} = ?$$

- What are the degrees of freedom achieved by lattice codes

$$\lim_{P \rightarrow \infty} \frac{R_L(P, \mathbf{H})}{\frac{1}{2} \log(P)} = ?$$

## Performance of Lattice Codes

$$R_L(P, \mathbf{h}, \mathbf{a}) \triangleq \frac{1}{2} \log \left( \frac{1 + P \|\mathbf{h}\|^2}{\|\mathbf{a}\|^2 + P(\|\mathbf{a}\|^2 \|\mathbf{h}\|^2 - (\mathbf{h}^T \mathbf{a})^2)} \right)$$

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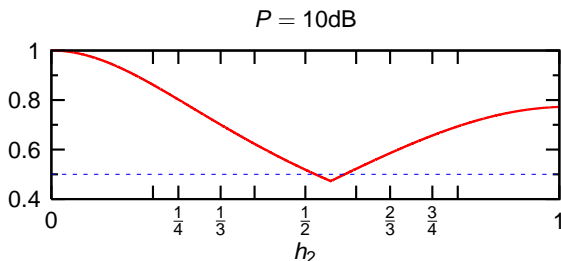
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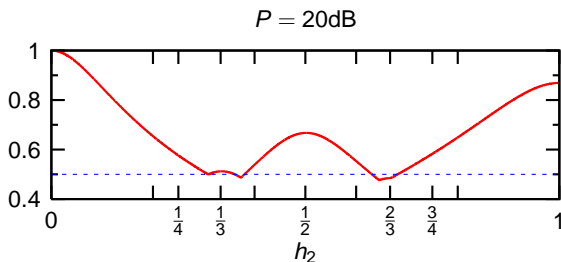
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$$\frac{\max_{\mathbf{a}} R_L(P, \mathbf{h}, \mathbf{a})}{\frac{1}{2} \log(1 + \|\mathbf{h}\|^2 P)}$$

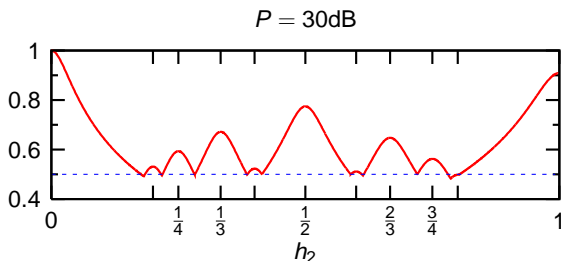
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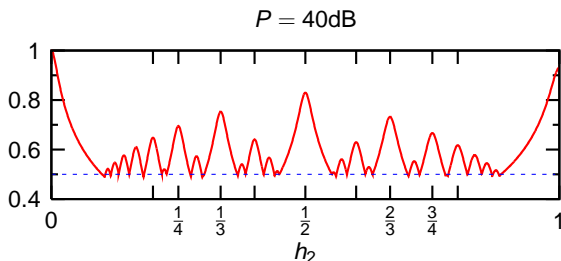


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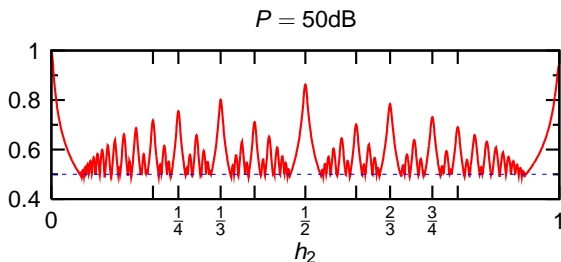
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- For  $A \approx P^{(K-1)/2K}$  and  $Q \approx P^{1/2K}$  satisfy power constraint and can remove noise

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Consider a simple interference channel without noise

$$y_1 = x_1 + x_2$$

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