

# Lattices in AWGN Networks: What's missing?

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Boston University

BIRS 2011: Algebraic Structure in Network Information Theory

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## *Some Disclaimers*

- This talk will include some **conjectures**...

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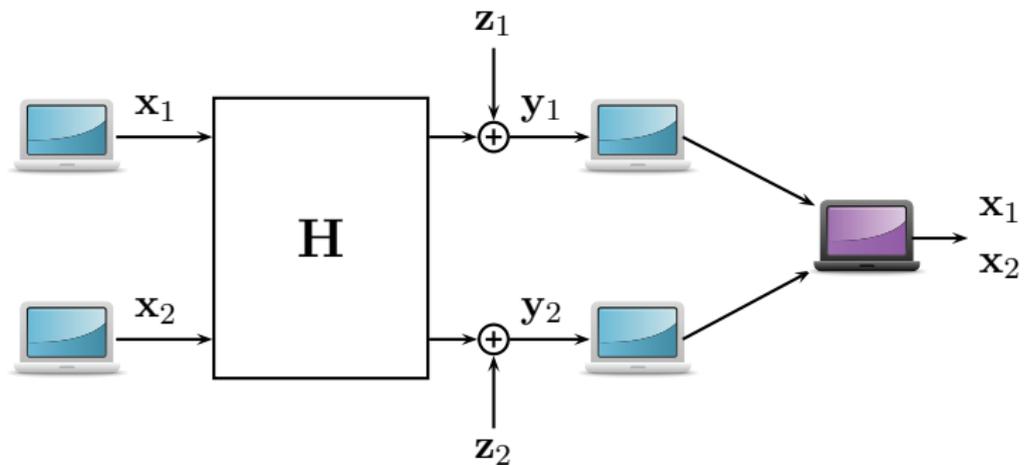
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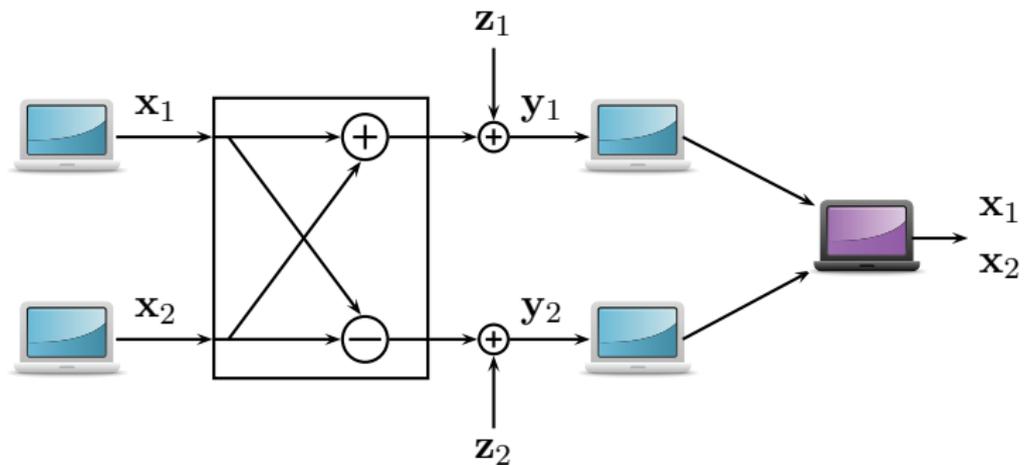
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- with the aim of making it a bit more interactive.
- Some of this will be tied to some (**local**) foods.
- ISIT '11 Tutorial co-taught with Michael Gastpar now available at ISIT website (and [iss.bu.edu/bobak](http://iss.bu.edu/bobak)).

## Compute-and-Forward



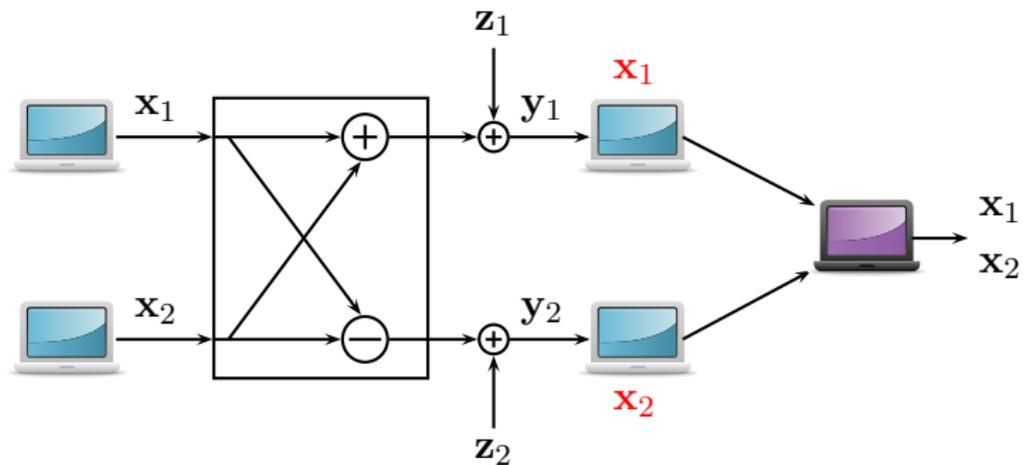
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## Compute-and-Forward



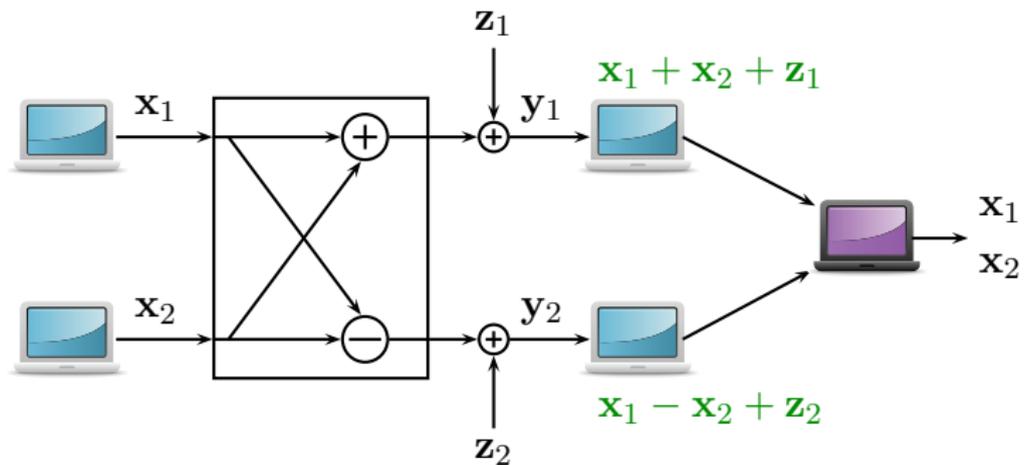
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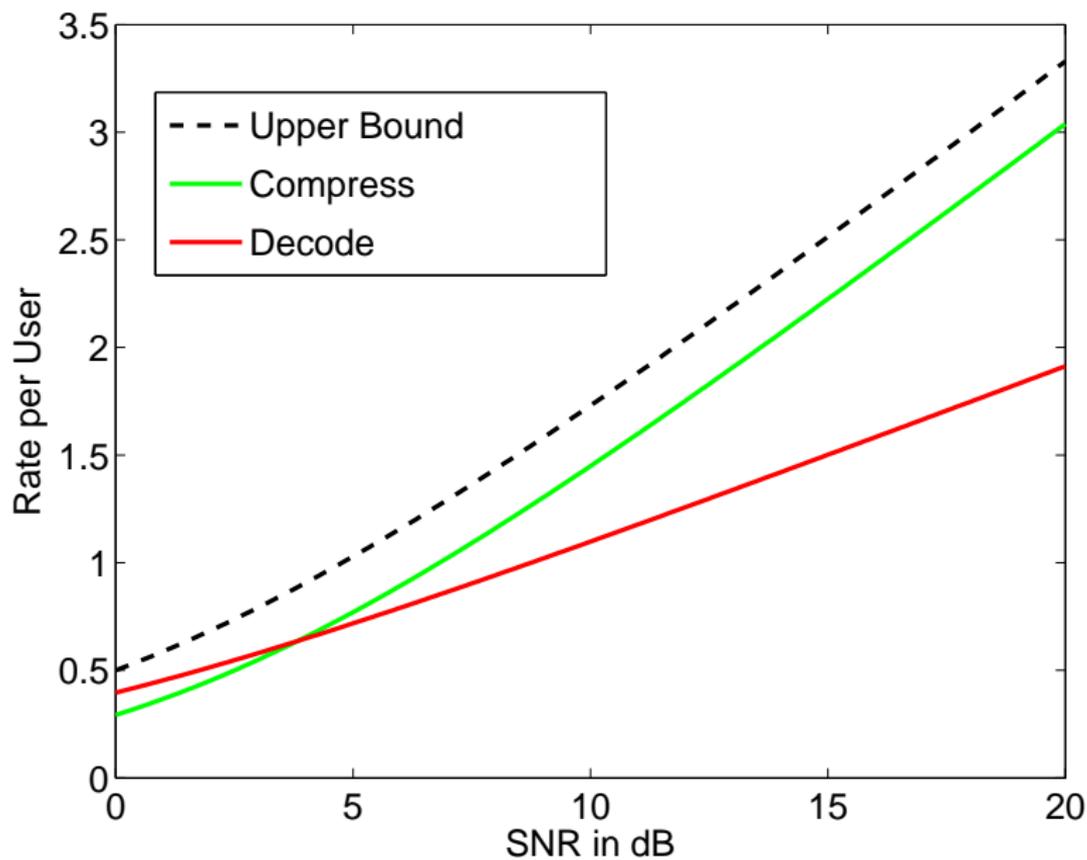
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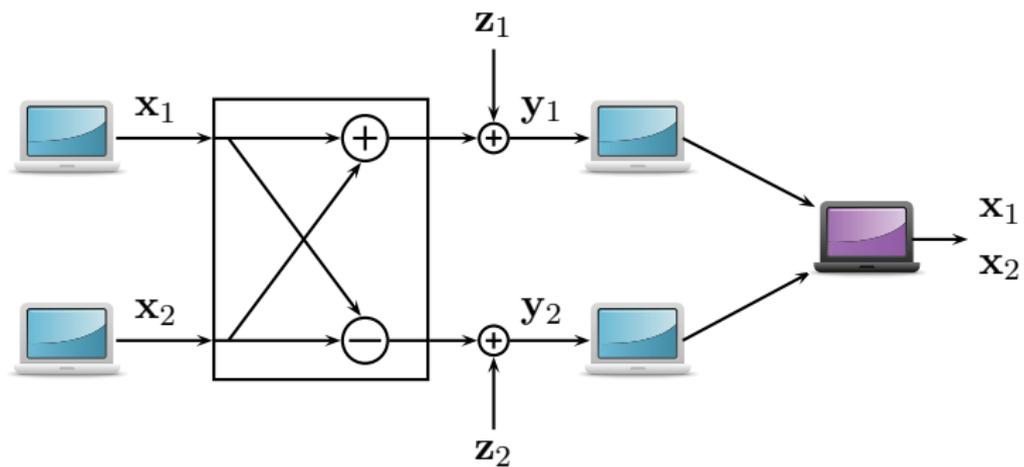


- Two users want to send messages across the network with the help of two relays.
- **Decode-and-Forward**: Each relay decodes one message.
- **Compress-and-Forward**: Relays send their observed signal to the destination without decoding.

## Compute-and-Forward

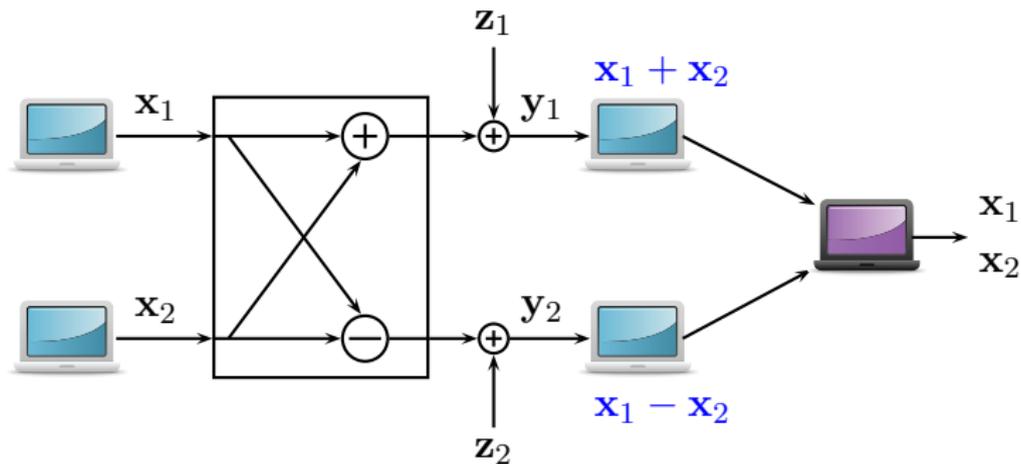


## Compute-and-Forward



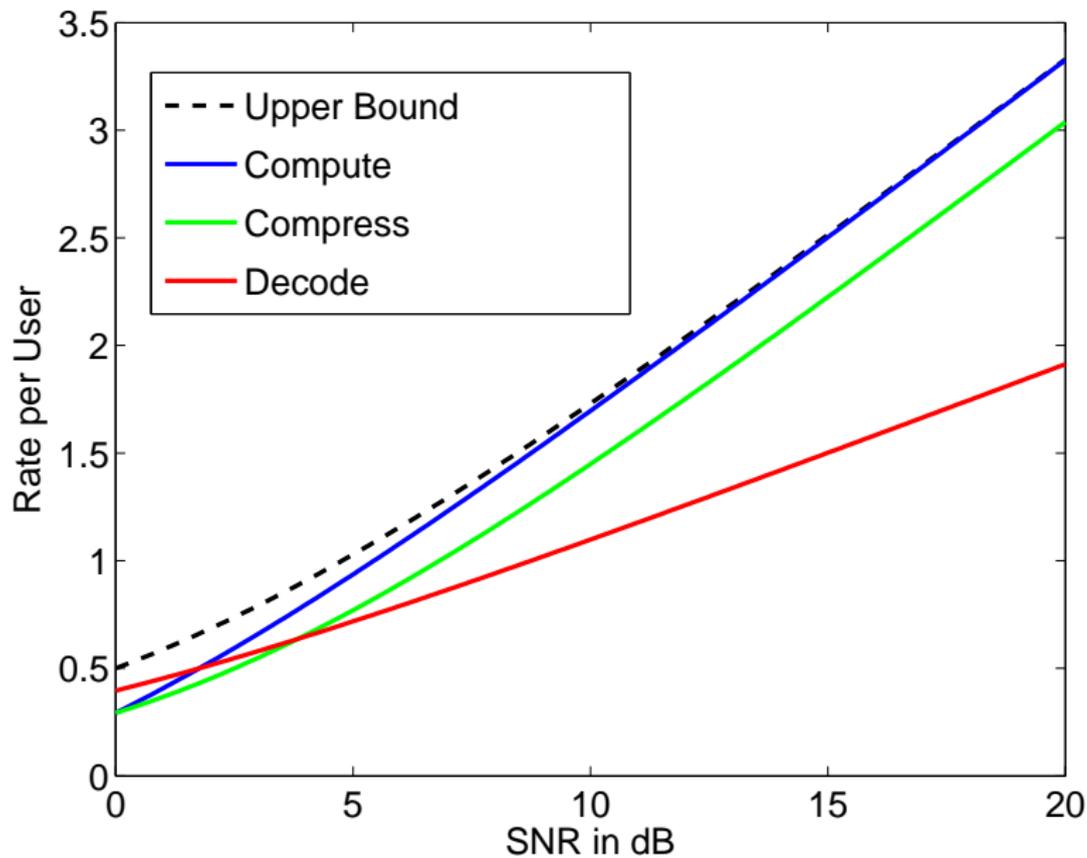
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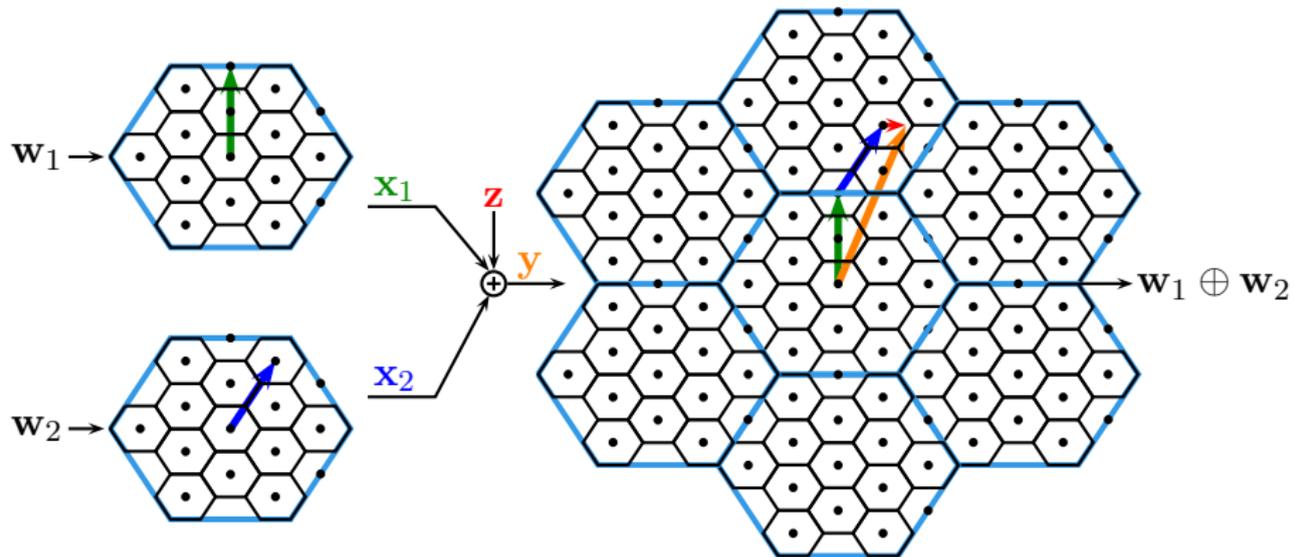


- What if each relay could decode a linear equation?
- **Compute-and-Forward:** One relay decodes the sum of codewords. Other relay decodes the difference.

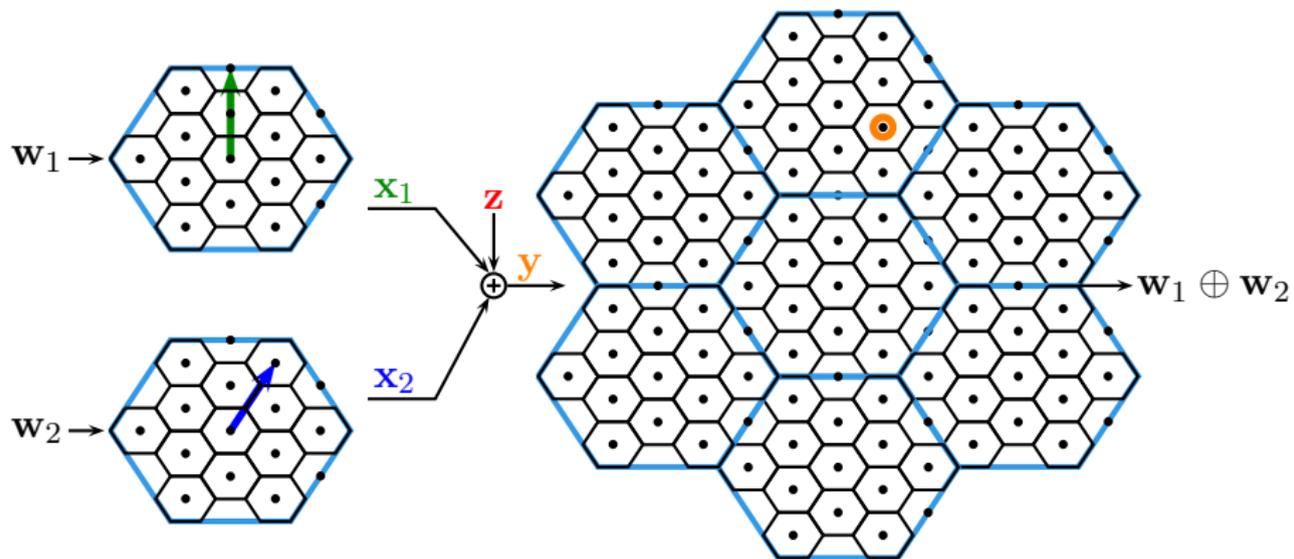
## Compute-and-Forward



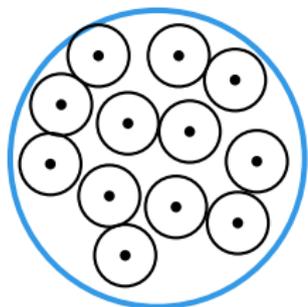
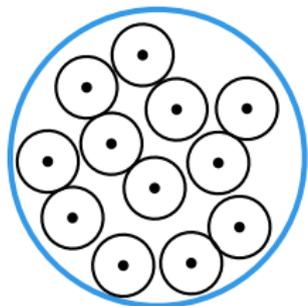
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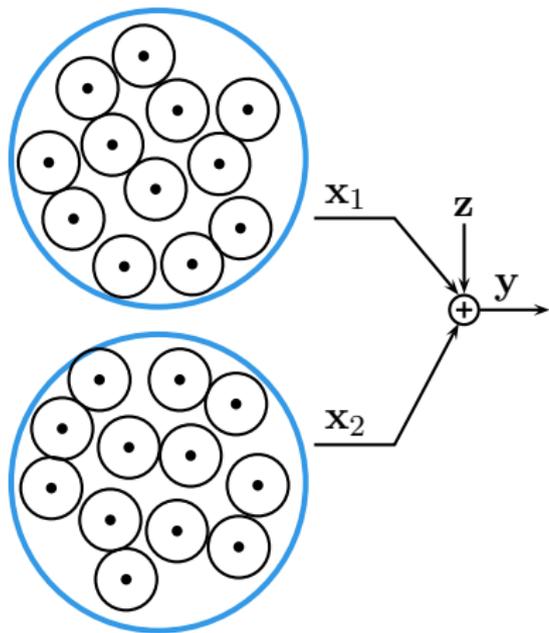
*Random i.i.d. codes are not good for computation*



$2^{nR}$  codewords each.

$2^{n2R}$  possible sums of codewords.

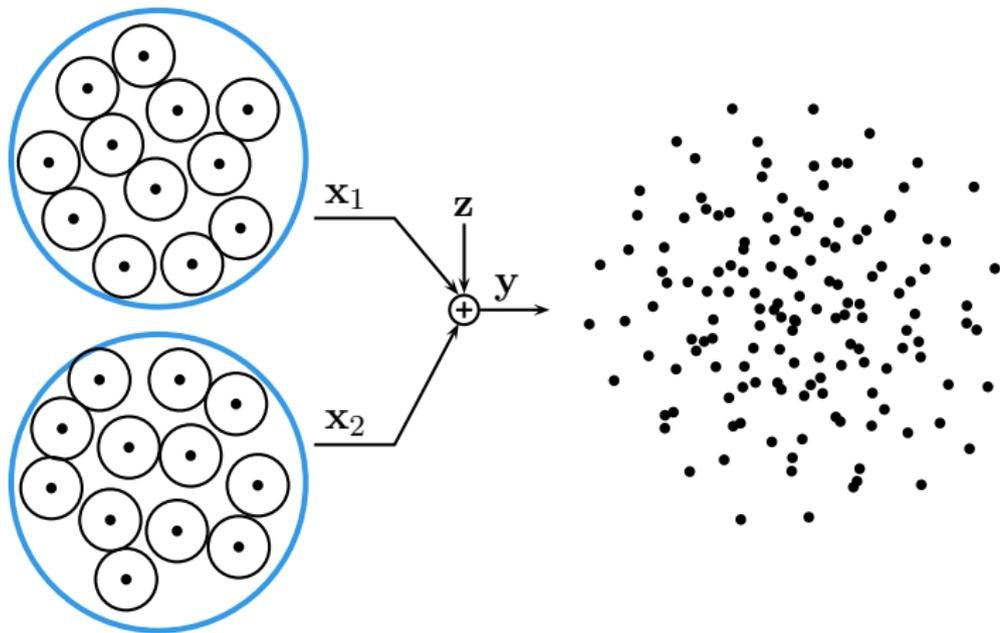
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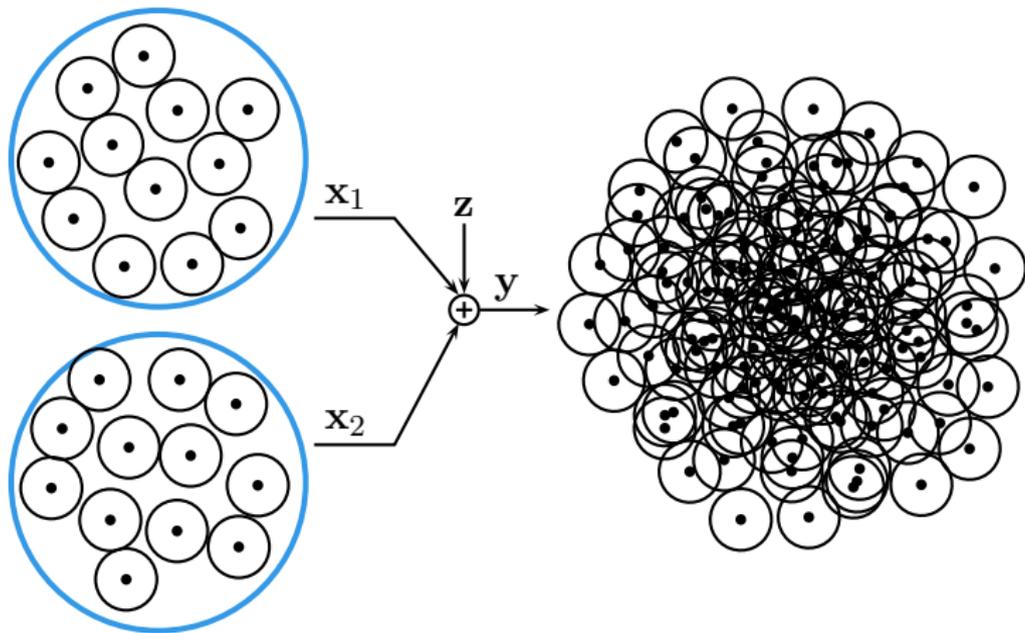
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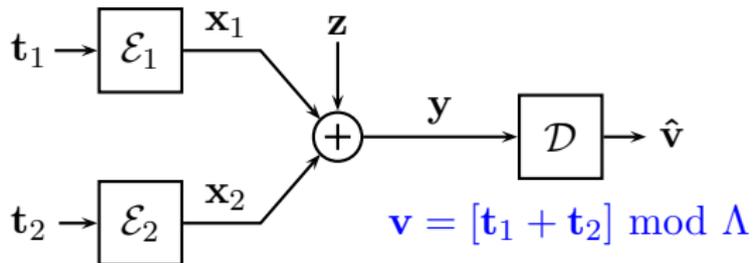
## Decoding the Sum of Lattice Codewords

Encoders use the same nested lattice codebook from **Erez-Zamir '04**.

Transmit lattice codewords:

$$\mathbf{x}_1 = \mathbf{t}_1$$

$$\mathbf{x}_2 = \mathbf{t}_2$$



Decoder **recovers modulo sum**.

$$\begin{aligned} & [\mathbf{y}] \bmod \Lambda \\ &= [\mathbf{x}_1 + \mathbf{x}_2 + \mathbf{z}] \bmod \Lambda \\ &= [\mathbf{t}_1 + \mathbf{t}_2 + \mathbf{z}] \bmod \Lambda \\ &= \left[ [\mathbf{t}_1 + \mathbf{t}_2] \bmod \Lambda + \mathbf{z} \right] \bmod \Lambda \quad \text{Distributive Law} \\ &= [\mathbf{v} + \mathbf{z}] \bmod \Lambda \end{aligned}$$

$$R = \frac{1}{2} \log \left( \frac{P}{N} \right)$$

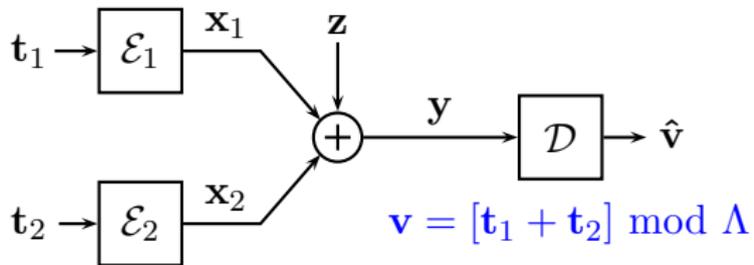
## Decoding the Sum of Lattice Codewords – MMSE Scaling

Encoders use the same nested lattice codebook from **Erez-Zamir '04**.

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$$\mathbf{x}_1 = [\mathbf{t}_1 + \mathbf{d}_1] \bmod \Lambda$$

$$\mathbf{x}_2 = [\mathbf{t}_2 + \mathbf{d}_2] \bmod \Lambda$$



Decoder scales by  $\alpha$ , removes dithers, **recovers modulo sum**.

$$[\alpha \mathbf{y} - \mathbf{d}_1 - \mathbf{d}_2] \bmod \Lambda$$

$$= [\alpha(\mathbf{x}_1 + \mathbf{x}_2 + \mathbf{z}) - \mathbf{d}_1 - \mathbf{d}_2] \bmod \Lambda$$

$$= [\mathbf{x}_1 + \mathbf{x}_2 - (1 - \alpha)(\mathbf{x}_1 + \mathbf{x}_2) + \alpha \mathbf{z} - \mathbf{d}_1 - \mathbf{d}_2] \bmod \Lambda$$

$$= \left[ [\mathbf{t}_1 + \mathbf{t}_2] \bmod \Lambda - (1 - \alpha)(\mathbf{x}_1 + \mathbf{x}_2) + \alpha \mathbf{z} \right] \bmod \Lambda$$

$$= \left[ \mathbf{v} - (1 - \alpha)(\mathbf{x}_1 + \mathbf{x}_2) + \alpha \mathbf{z} \right] \bmod \Lambda$$

**Effective Noise**  $N_{\text{EFFEC}} = (1 - \alpha)^2 2P + \alpha^2 N$

## Decoding the Sum of Lattice Codewords – MMSE Scaling

- Effective noise after scaling is  $N_{\text{EFFEC}} = (1 - \alpha)^2 2P + \alpha^2 N$ .
- Minimized by setting  $\alpha$  to be the **MMSE coefficient**:

$$\alpha_{\text{MMSE}} = \frac{2P}{N + 2P}$$

- Plugging in, we get

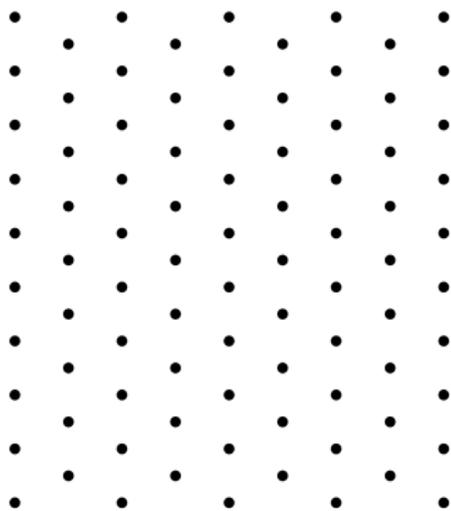
$$N_{\text{EFFEC}} = \frac{2NP}{N + 2P}$$

- Resulting rate is

$$R = \frac{1}{2} \log \left( \frac{P}{N_{\text{EFFEC}}} \right) = \frac{1}{2} \log \left( \frac{1}{2} + \frac{P}{N} \right)$$

- What happened to the “one plus” term?

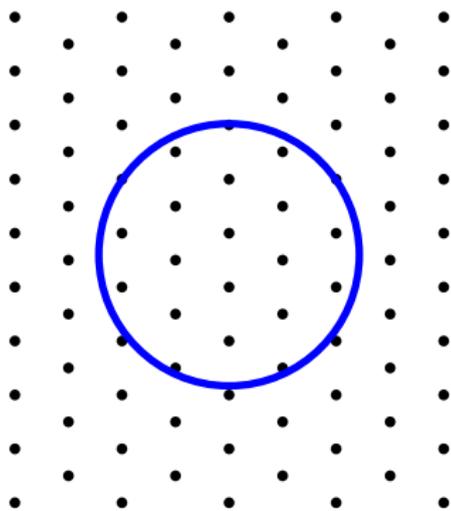
## Where is the “one plus”?



Set fine lattice density using:

$$R = \frac{1}{2} \log \left( 1 + \frac{P}{N} \right)$$

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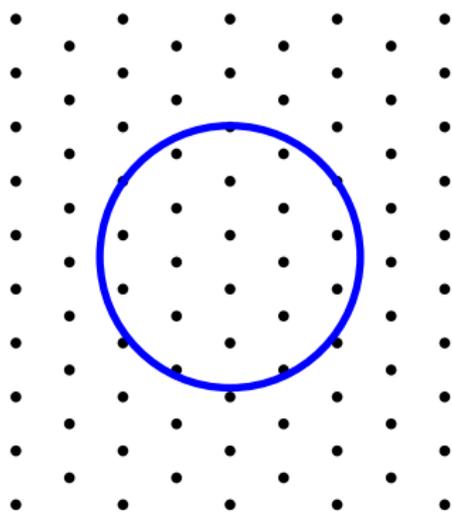


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Cut out codebook at **power  $P$** .

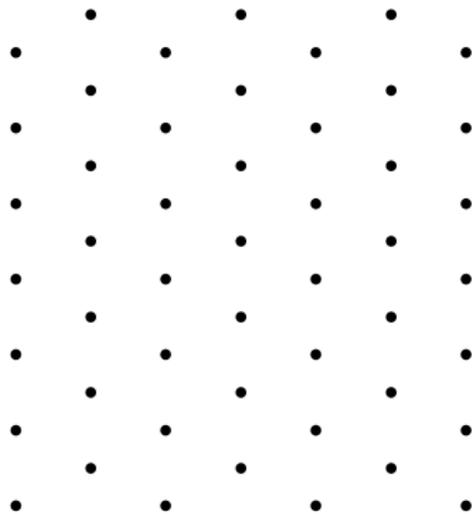
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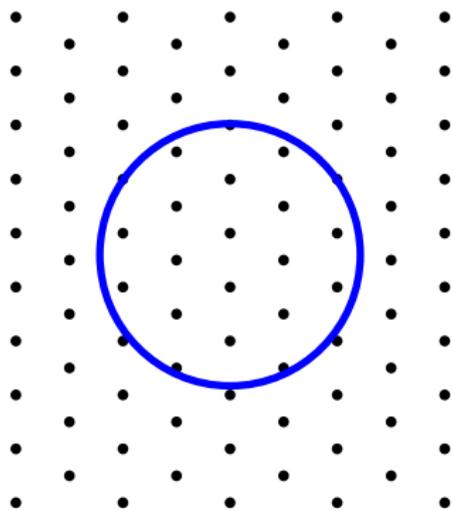
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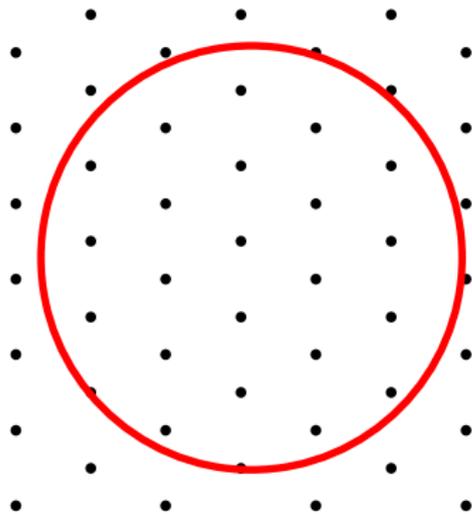
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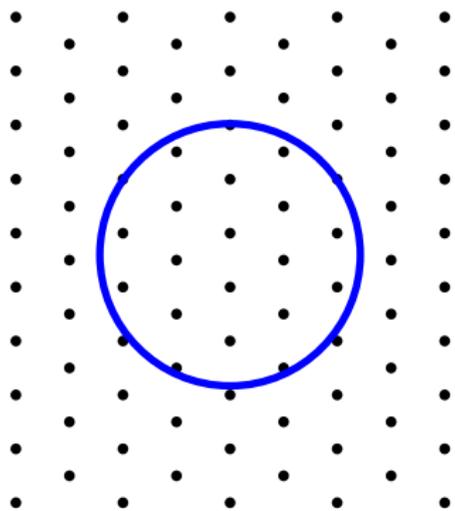
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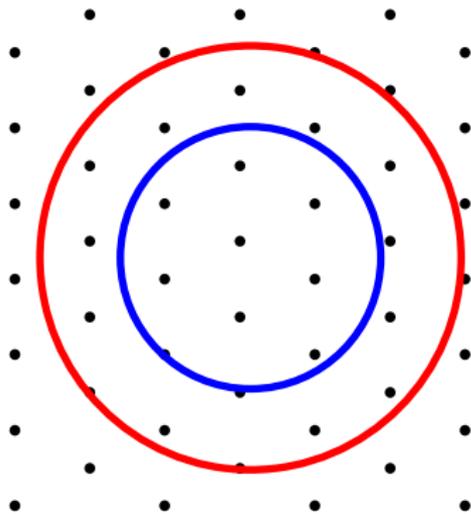
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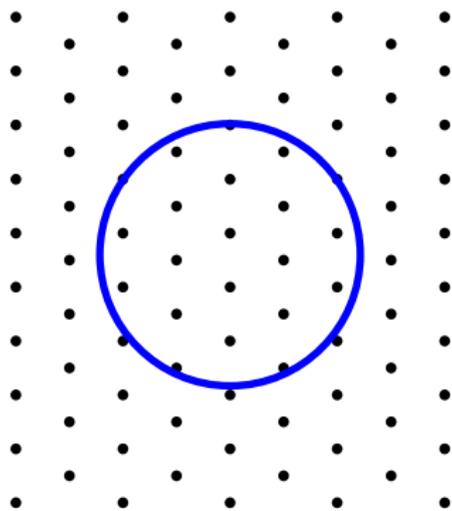


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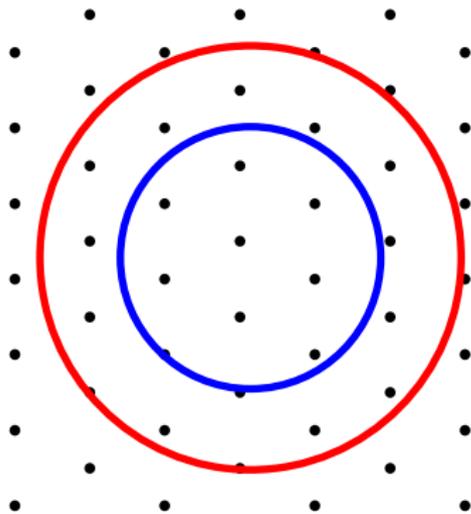
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Resulting codebook only has  $R = \frac{1}{2} \log \left( \frac{1}{2} + \frac{P}{N} \right)!$

## Where is the “one plus”?

- This limitation seems inherent to nested lattice codes combined with lattice decoding.
- Also seems inherent to any scheme that treats all codewords as “the same” (e.g. ML decoding). Connected to decoding analysis in **Wilson-Narayanan-Pfister-Sprintson '10**.
- What about MAP decoding?
- **Ice Wine Problem**: Prove that the sum of codewords  $\mathbf{x}_1 + \mathbf{x}_2$  can be recovered from  $\mathbf{y} = \mathbf{x}_1 + \mathbf{x}_2 + \mathbf{z}$  at rate

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Or prove that this is impossible.

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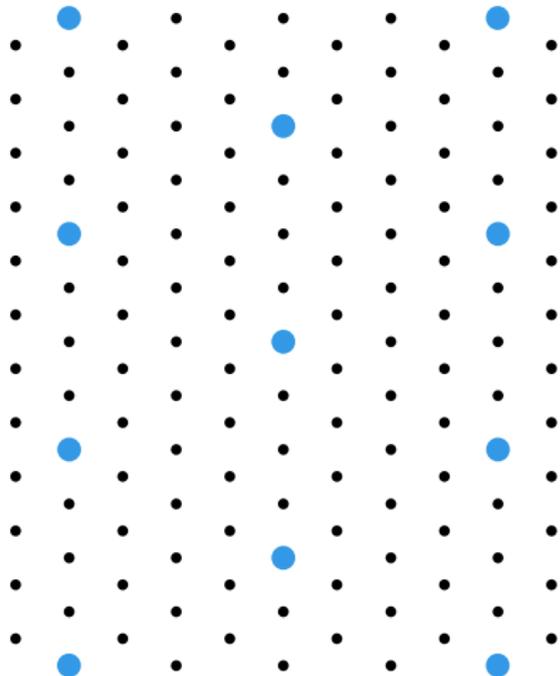
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- Loss is at most  $1/2$  of a bit. Is this such a big deal? **Yes, especially for layering.**
- Consider employing a superposition of many lattice codewords.
- As the number of layers increases, the effective SNR of each layer decreases.
- Hard to analyze layered lattice codebooks outside the high SNR regime.
- Shows up in interference channels (e.g. **Sridharan et al. '08**)

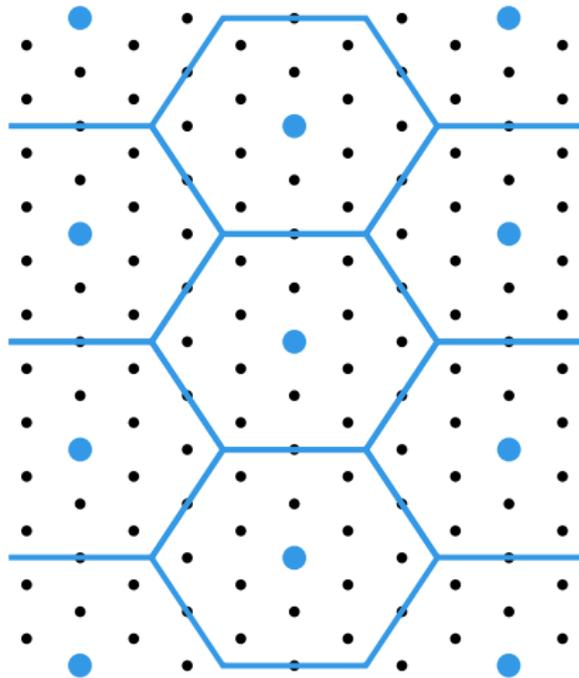
## Unequal Power Constraints – Double Nesting

- What if the power constraints are not equal?
- Idea from **Nam-Chung-Lee '10**:
- Draw the codewords from the **same fine lattice**  $\Lambda_{\text{FINE}}$ .
- Use two nested coarse lattices  $\Lambda_1$  and  $\Lambda_2$  to enforce the power constraints  $P_1$  and  $P_2$ .



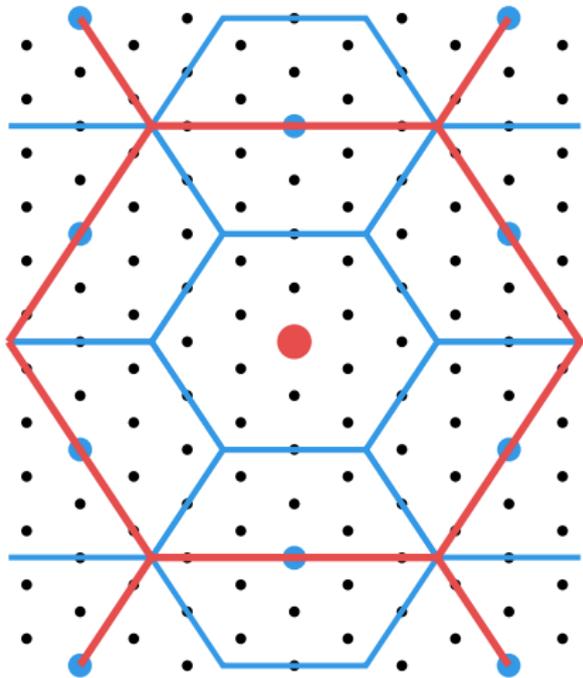
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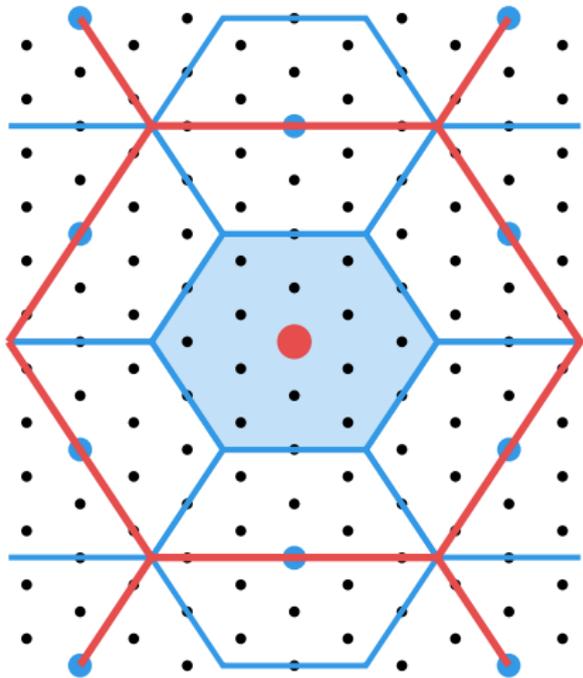
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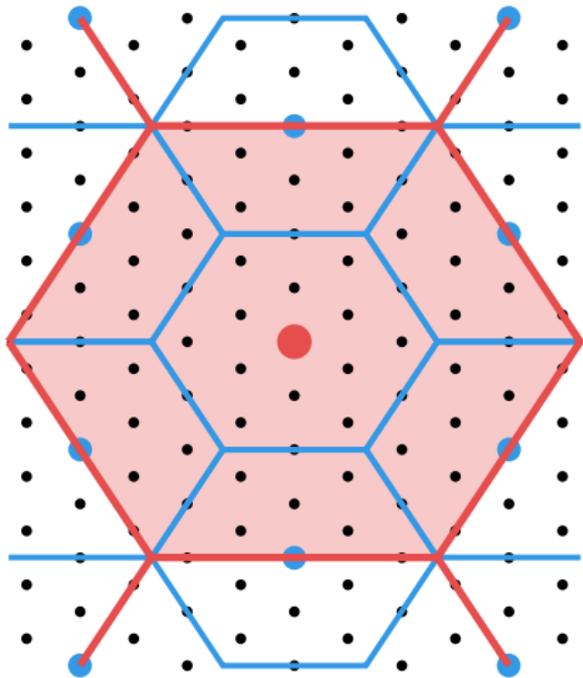
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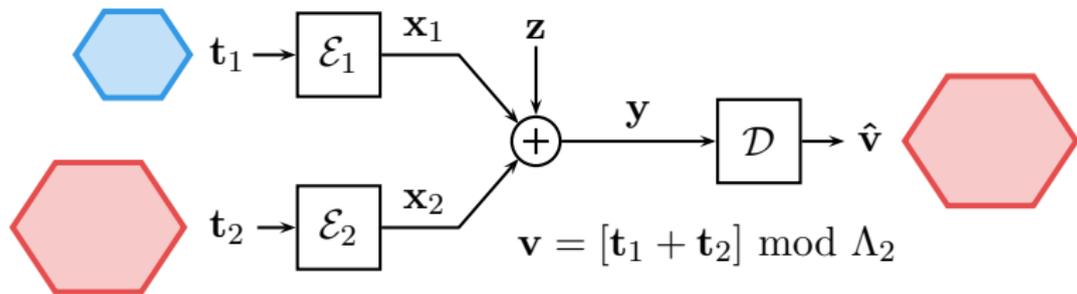


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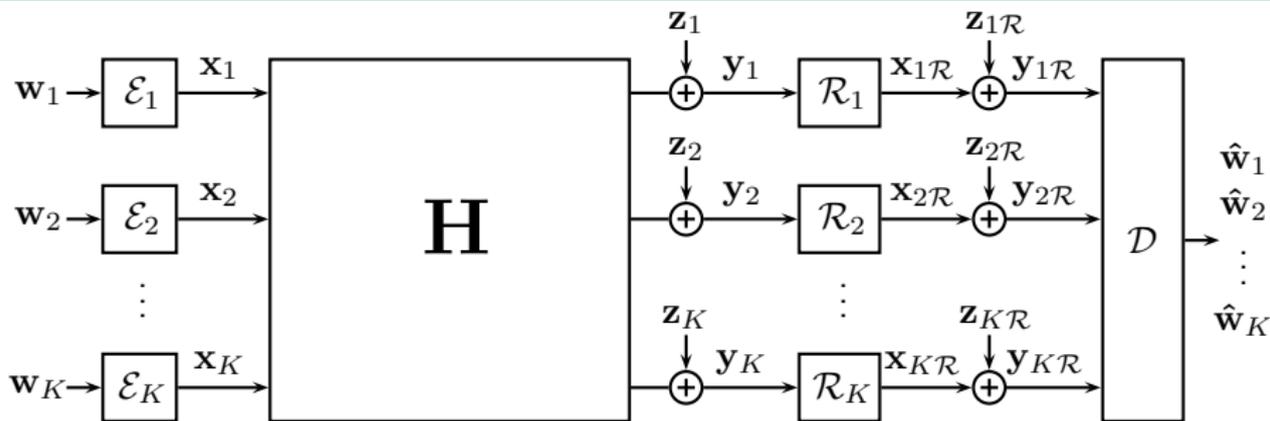


- Encoder 1 sends  $\mathbf{x}_1 = [\mathbf{t}_1 + \mathbf{d}_1] \bmod \Lambda_1$ . Coarse lattice  $\Lambda_1$  has second moment  $P_1$ .
- Encoder 2 sends  $\mathbf{x}_2 = [\mathbf{t}_2 + \mathbf{d}_2] \bmod \Lambda_2$ . Coarse lattice  $\Lambda_2$  has second moment  $P_2 > P_1$ .
- Decoder performs MMSE scaling, remove dithers, recovers  $\bmod \Lambda_2$  sum.

$$R_1 = \frac{1}{2} \log \left( \frac{P_1}{P_1 + P_2} + \frac{P_1}{N} \right)$$

$$R_2 = \frac{1}{2} \log \left( \frac{P_2}{P_1 + P_2} + \frac{P_2}{N} \right)$$

## Case Study – Hadamard Relay Network



- Equal rates  $R$ .  $\mathbf{H}$  is a Hadamard matrix,  $\mathbf{H}\mathbf{H}^T = K\mathbf{I}$

Upper Bound

$$\frac{1}{2} \log \left( 1 + \frac{P}{N} \right)$$

Compress-and-Forward

$$\frac{1}{2} \log \left( 1 + \frac{P}{N} \frac{P}{N + KP} \right)$$

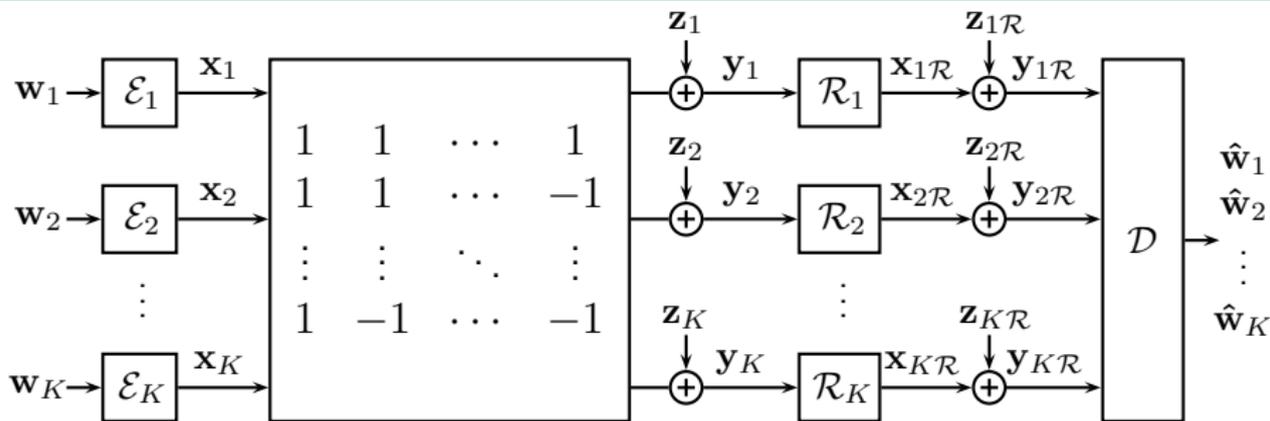
Compute-and-Forward

$$\frac{1}{2} \log \left( \frac{1}{K} + \frac{P}{N} \right)$$

Decode-and-Forward

$$\frac{1}{2K} \log \left( 1 + \frac{KP}{N} \right)$$

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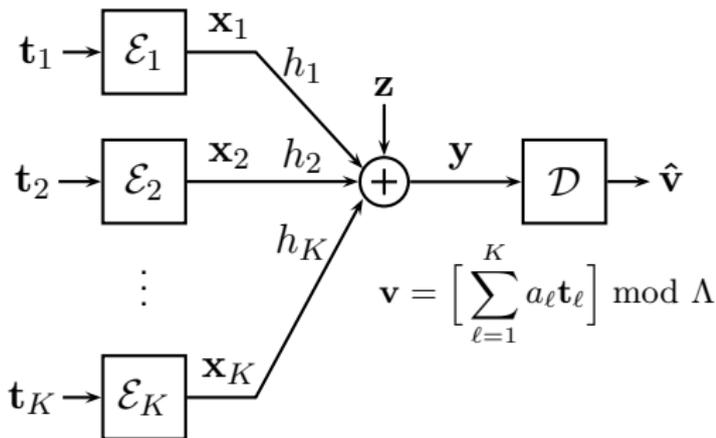
## Compute-and-Forward: Fading Channels

Transmitters **do not know** channel realization.

Encoders use the same nested lattice codebook.

Transmit dithered codewords:

$$\mathbf{x}_\ell = [\mathbf{t}_\ell + \mathbf{d}_\ell] \bmod \Lambda$$



- Decoder removes dithers and recovers **integer combination**

$$\mathbf{v} = \left[ \sum_{\ell=1}^K a_\ell \mathbf{t}_\ell \right] \bmod \Lambda$$

- Receiver can use its knowledge of the channel gains to match the equation coefficients  $a_\ell$  to the channel coefficients  $h_\ell$ .

- **Distributive Law** also holds for integer combinations. Let  $a, b \in \mathbb{Z}$ .

$$\begin{aligned} & \left[ a[\mathbf{x}_1] \bmod \Lambda + b[\mathbf{x}_2] \bmod \Lambda \right] \bmod \Lambda \\ &= \left[ a\left(\mathbf{x}_1 - Q_\Lambda(\mathbf{x}_1)\right) + b\left(\mathbf{x}_2 - Q_\Lambda(\mathbf{x}_2)\right) \right] \bmod \Lambda \\ &= \left[ a\mathbf{x}_1 + b\mathbf{x}_2 - aQ_\Lambda(\mathbf{x}_1) - bQ_\Lambda(\mathbf{x}_2) \right] \bmod \Lambda \\ &= [a\mathbf{x}_1 + b\mathbf{x}_2] \bmod \Lambda \end{aligned}$$

- Last step follows since since  $aQ_\Lambda(\mathbf{x}_1)$  and  $bQ_\Lambda(\mathbf{x}_2)$  are elements of the lattice  $\Lambda$ .

## Compute-and-Forward: Fading Channels

- Transmit dithered codewords  $\mathbf{x}_\ell = [\mathbf{t}_\ell + \mathbf{d}_\ell] \bmod \Lambda$
- Decoder removes dithers and recovers integer combination

$$\begin{aligned} & \left[ \mathbf{y} - \sum_{\ell=1}^K a_\ell \mathbf{d}_\ell \right] \bmod \Lambda \\ &= \left[ \sum_{\ell=1}^K h_\ell \mathbf{x}_\ell + \mathbf{z} - \sum_{\ell=1}^K a_\ell \mathbf{d}_\ell \right] \bmod \Lambda \\ &= \left[ \sum_{\ell=1}^K a_\ell (\mathbf{x}_\ell - \mathbf{d}_\ell) + \sum_{\ell=1}^K (h_\ell - a_\ell) \mathbf{x}_\ell + \mathbf{z} \right] \bmod \Lambda \\ &= \left[ \left[ \sum_{\ell=1}^K a_\ell \mathbf{t}_\ell \right] \bmod \Lambda + \underbrace{\sum_{\ell=1}^K (h_\ell - a_\ell) \mathbf{x}_\ell + \mathbf{z}}_{\text{Effective Noise}} \right] \bmod \Lambda \quad \text{Distributive Law} \end{aligned}$$

## Compute-and-Forward: Fading Channels – Effective Noise

- Effective noise due to **mismatch** between channel coefficients  $\mathbf{h} = [h_1 \cdots h_K]^T$  and equation coefficients  $\mathbf{a} = [a_1 \cdots a_K]^T$ .

$$N_{\text{EFFEC}} = N + P\|\mathbf{h} - \mathbf{a}\|^2$$
$$R = \frac{1}{2} \log \left( \frac{P}{N + P\|\mathbf{h} - \mathbf{a}\|^2} \right)$$

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- Can do better with **MMSE scaling**.

$$N_{\text{EFFEC}} = \alpha^2 N + P\|\alpha\mathbf{h} - \mathbf{a}\|^2$$
$$R = \max_{\alpha} \frac{1}{2} \log \left( \frac{P}{\alpha^2 N + P\|\alpha\mathbf{h} - \mathbf{a}\|^2} \right)$$
$$= \frac{1}{2} \log \left( \frac{N + P\|\mathbf{h}\|^2}{N\|\mathbf{a}\|^2 + P(\|\mathbf{h}\|^2\|\mathbf{a}\|^2 - (\mathbf{h}^T \mathbf{a})^2)} \right)$$

- See **Nazer-Gastpar '11** for more details.

- The rate expression simplifies in some special cases.

$$R = \frac{1}{2} \log \left( \frac{N + P\|\mathbf{h}\|^2}{N\|\mathbf{a}\|^2 + P(\|\mathbf{h}\|^2\|\mathbf{a}\|^2 - (\mathbf{h}^T \mathbf{a})^2)} \right)$$

- Integer channels:  $\mathbf{h} = \mathbf{a}$ .

$$R = \frac{1}{2} \log \left( \frac{1}{\|\mathbf{a}\|^2} + \frac{P}{N} \right)$$

- Recovering a single message: Set  $\mathbf{a} = \delta_m$ , the  $m^{\text{th}}$  unit vector.

$$R = \frac{1}{2} \log \left( 1 + \frac{h_m^2 P}{N + P \sum_{\ell \neq m} h_\ell^2} \right)$$

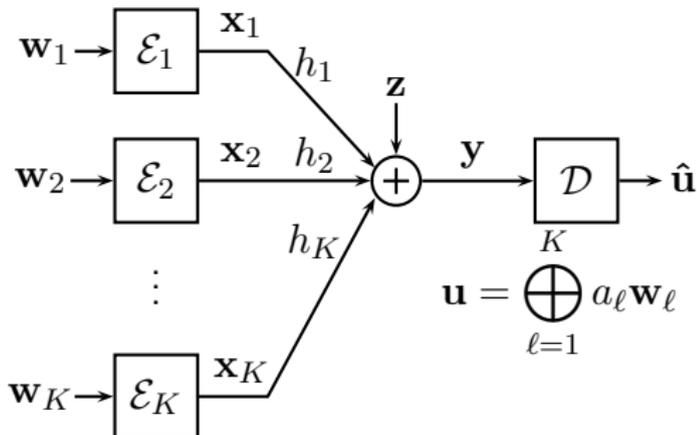
## Compute-and-Forward: Fading Channels – Finite Field Message

Transmitters **do not know** channel realization.

Encoders use the same nested lattice codebook.

Transmit dithered codewords:

$$\mathbf{x}_\ell = [\mathbf{t}_\ell + \mathbf{d}_\ell] \bmod \Lambda$$



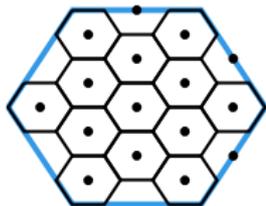
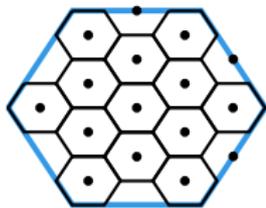
- Due to Construction A, mapping  $\mathbf{t}_\ell = \phi(\mathbf{w}_\ell)$  between messages and lattice points **preserves linearity**.

$$\phi^{-1}\left(\left[\sum_{\ell=1}^K a_\ell \mathbf{t}_\ell\right] \bmod \Lambda\right) = \left[\sum_{\ell=1}^K a_\ell \mathbf{w}_\ell\right] \bmod q = \bigoplus_{\ell=1}^K a_\ell \mathbf{w}_\ell$$

- Digital interface that fits well with **network coding**.

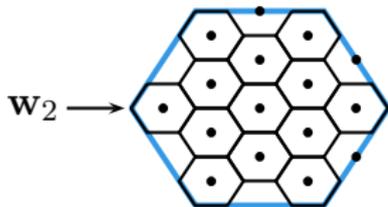
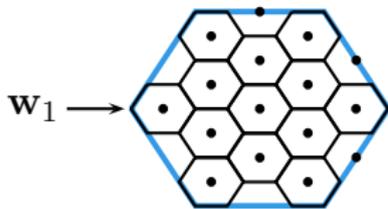
## Compute-and-Forward: Fading Channels – Illustration

All users pick the **same nested lattice code**:



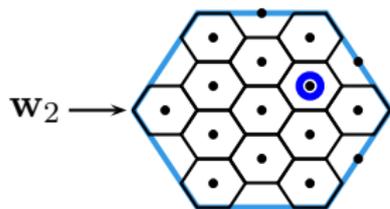
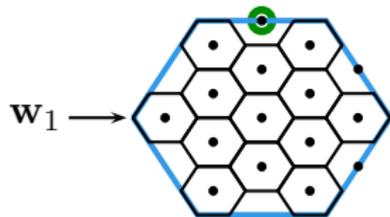
## Compute-and-Forward: Fading Channels – Illustration

Choose messages over field  $\mathbf{w}_\ell \in \mathbb{F}_q^k$ :



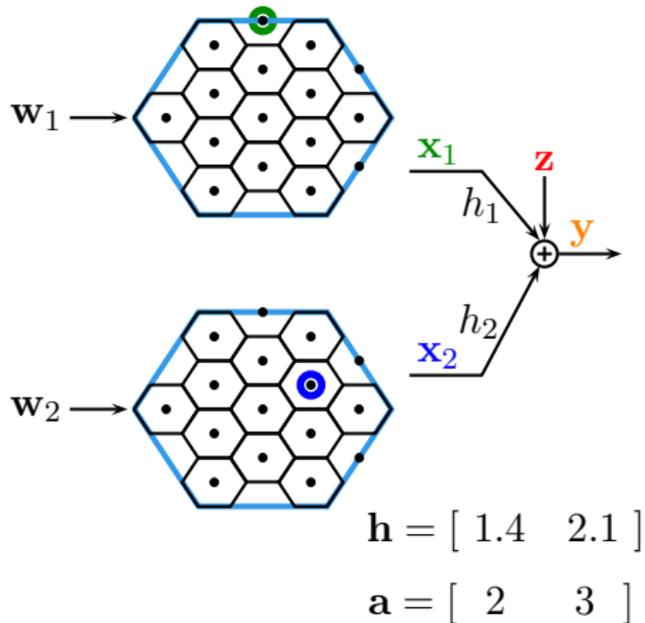
## Compute-and-Forward: Fading Channels – Illustration

Map  $\mathbf{w}_\ell$  to lattice point  $\mathbf{t}_\ell = \phi(\mathbf{w}_\ell)$ :



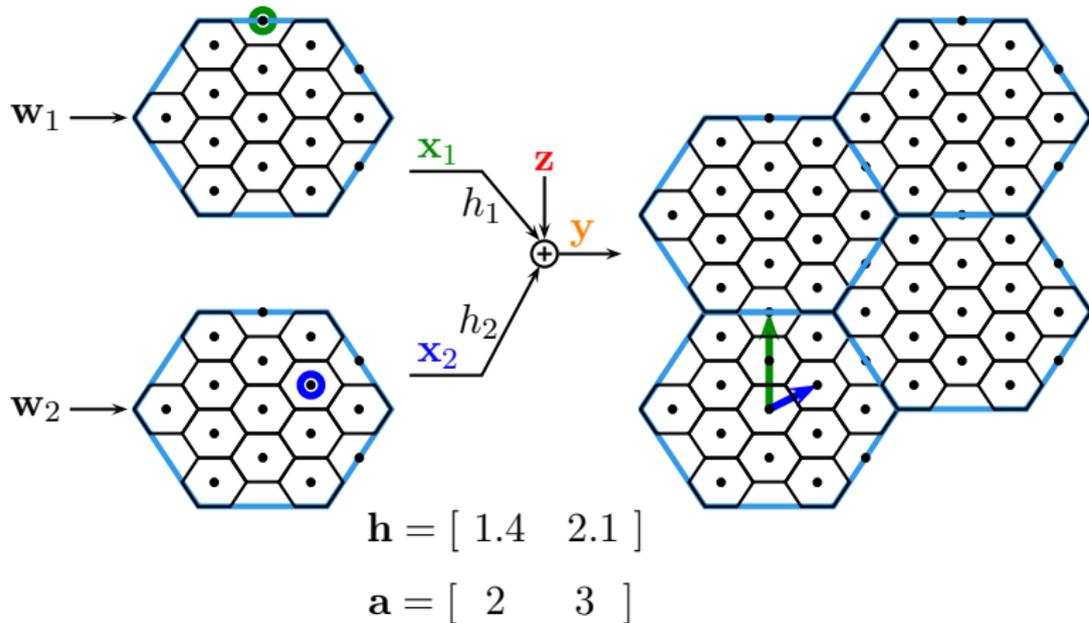
## Compute-and-Forward: Fading Channels – Illustration

Transmit lattice points over the channel:



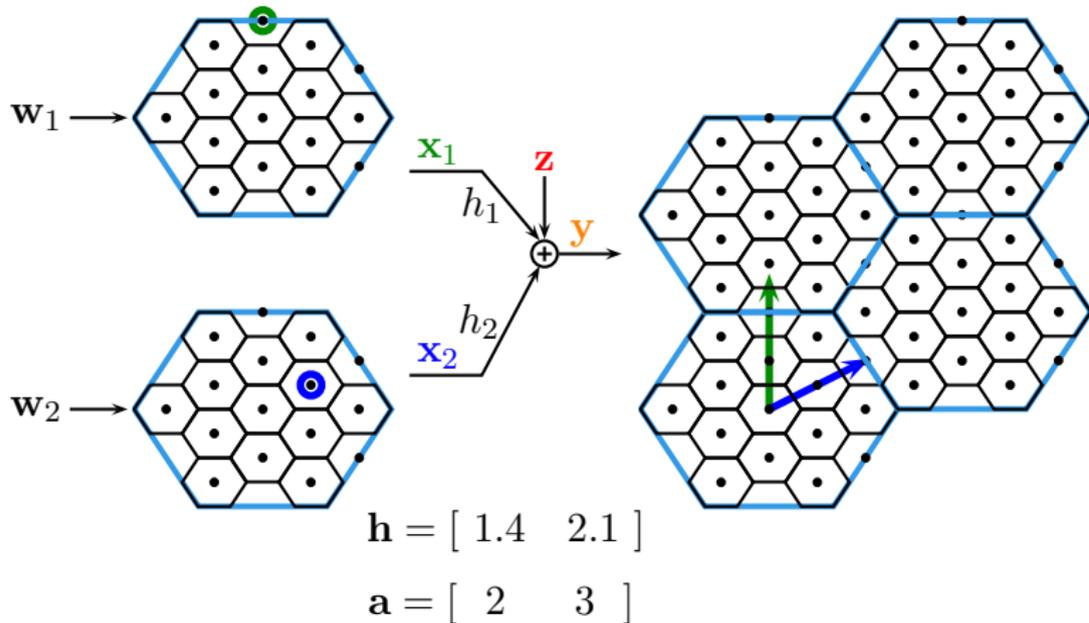
## Compute-and-Forward: Fading Channels – Illustration

Transmit lattice points over the channel:



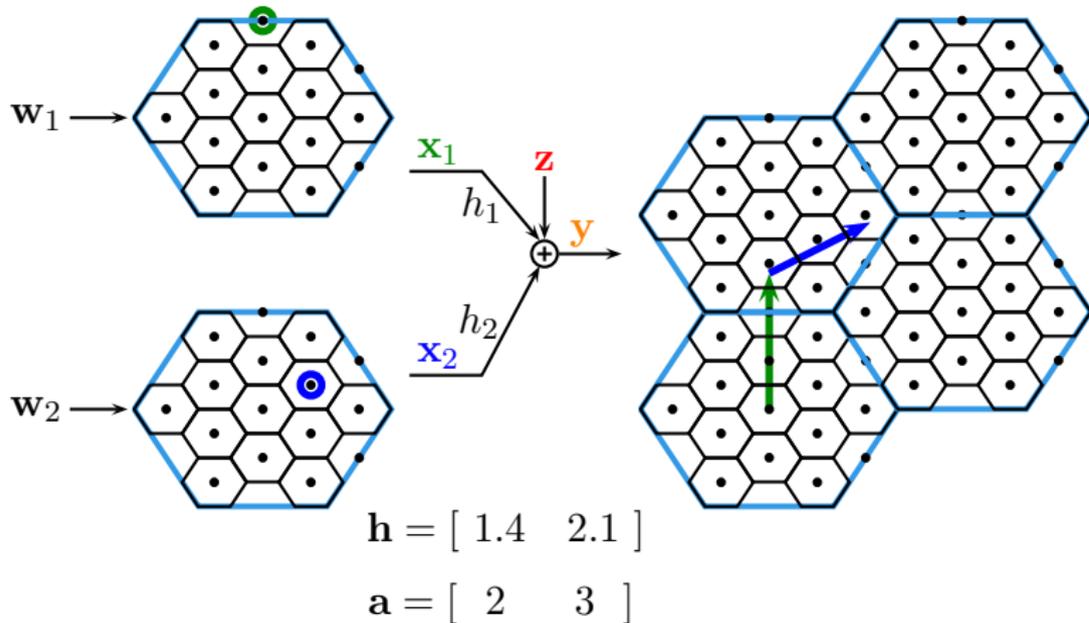
## Compute-and-Forward: Fading Channels – Illustration

Lattice codewords are scaled by channel coefficients:



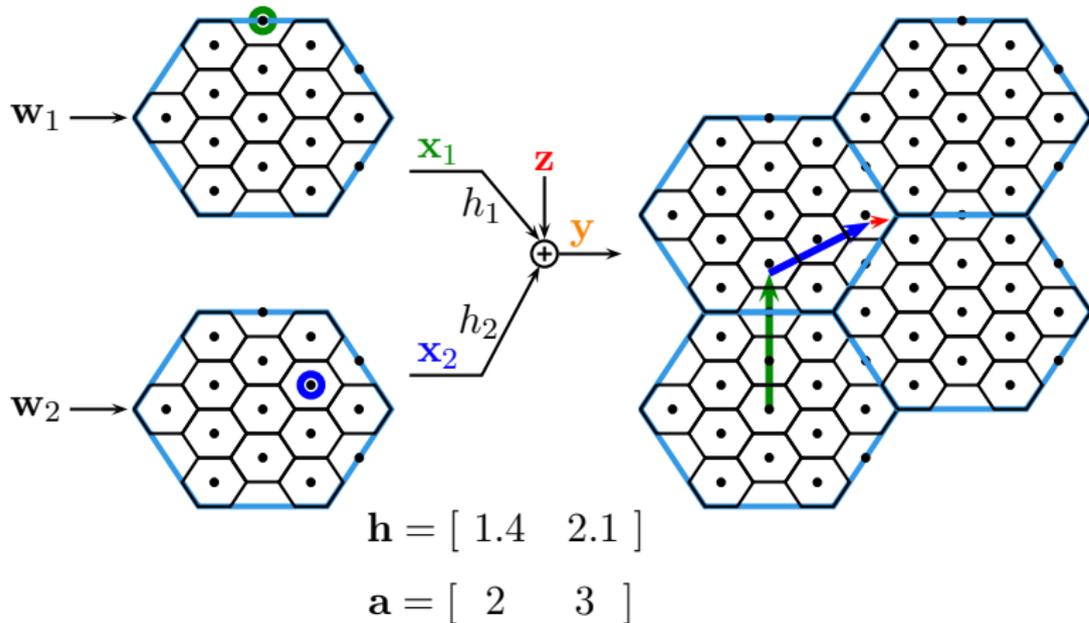
## Compute-and-Forward: Fading Channels – Illustration

Scaled codewords added together plus **noise**:



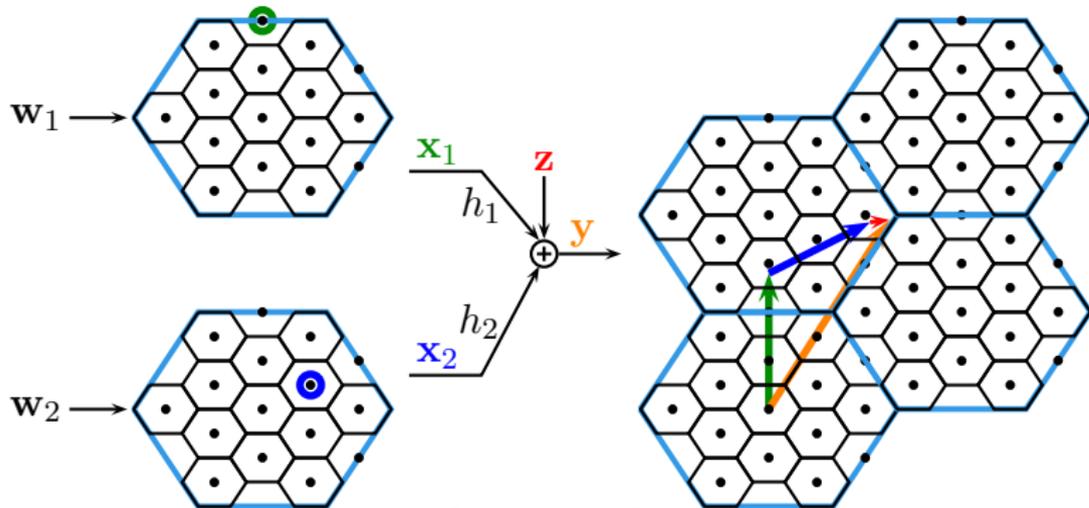
## Compute-and-Forward: Fading Channels – Illustration

Scaled codewords added together plus **noise**:



## Compute-and-Forward: Fading Channels – Illustration

Extra noise penalty for non-integer channel coefficients:



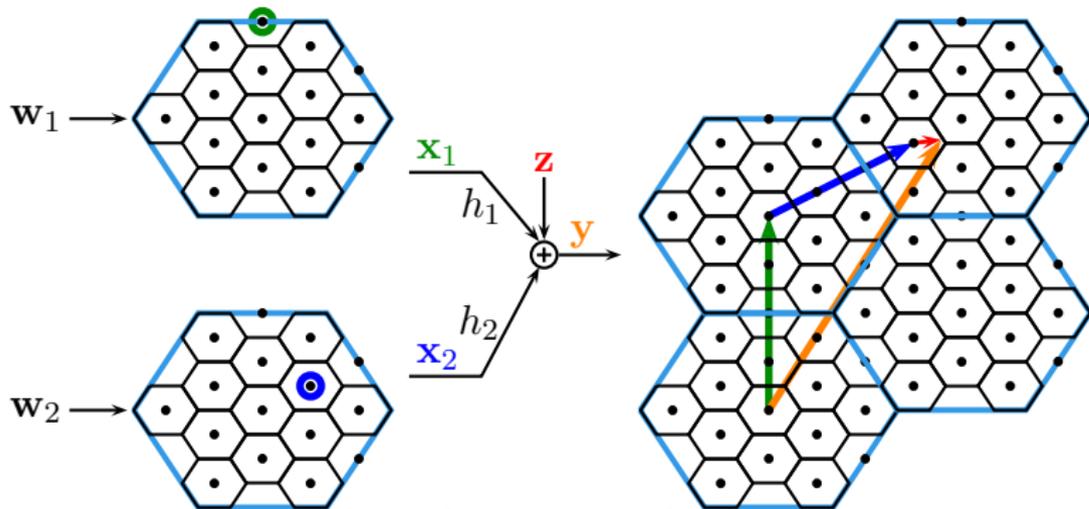
$$\mathbf{h} = [1.4 \ 2.1]$$

$$\mathbf{a} = [2 \ 3]$$

$$\text{Effective noise: } N + P\|\mathbf{h} - \mathbf{a}\|^2$$

# Compute-and-Forward: Fading Channels – Illustration

Scale output by  $\alpha$  to reduce non-integer noise penalty:



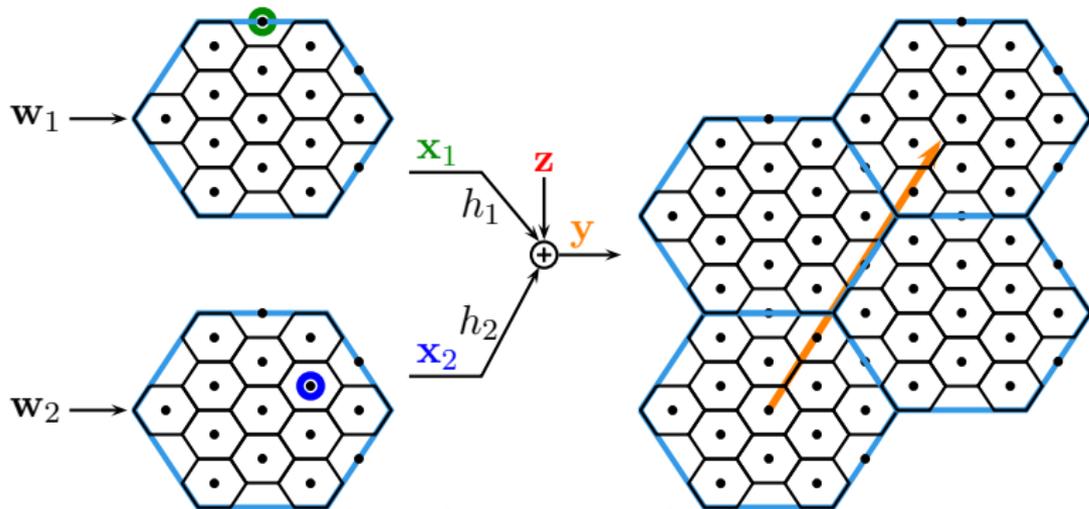
$$\alpha \mathbf{h} = [ \alpha 1.4 \quad \alpha 2.1 ]$$

$$\mathbf{a} = [ 2 \quad 3 ]$$

$$\text{Effective noise: } \alpha^2 N + P \|\alpha \mathbf{h} - \mathbf{a}\|^2$$

## Compute-and-Forward: Fading Channels – Illustration

Scale output by  $\alpha$  to reduce non-integer noise penalty:



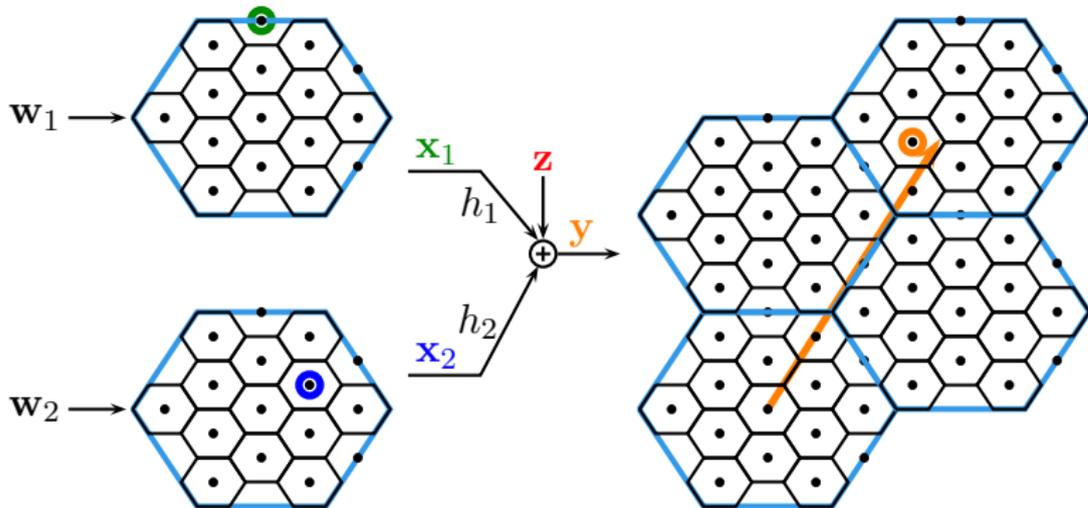
$$\alpha \mathbf{h} = [ \alpha 1.4 \quad \alpha 2.1 ]$$

$$\mathbf{a} = [ 2 \quad 3 ]$$

$$\text{Effective noise: } \alpha^2 N + P \|\alpha \mathbf{h} - \mathbf{a}\|^2$$

# Compute-and-Forward: Fading Channels – Illustration

Decode to closest lattice point:



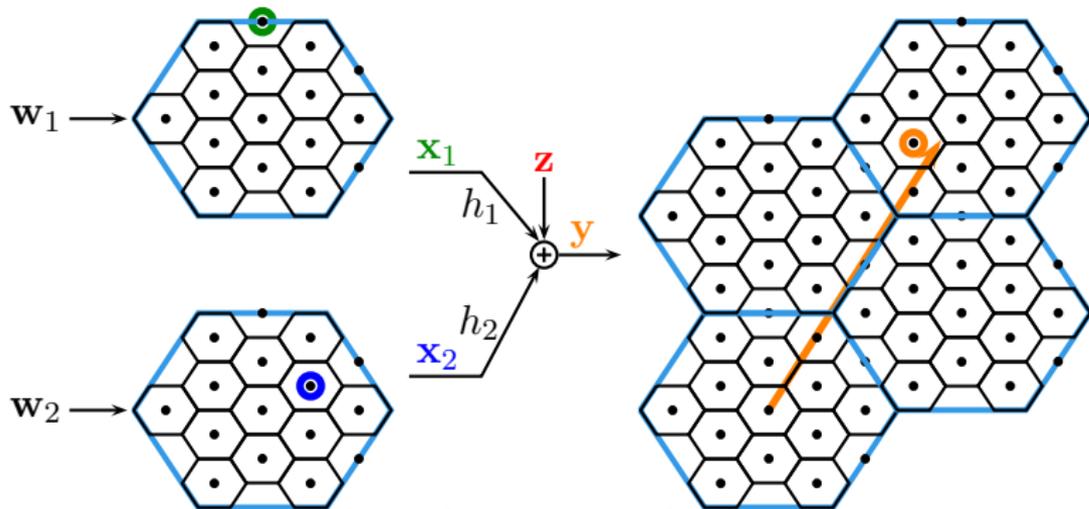
$$\alpha \mathbf{h} = [ \alpha 1.4 \quad \alpha 2.1 ]$$

$$\mathbf{a} = [ 2 \quad 3 ]$$

$$\text{Effective noise: } \alpha^2 N + P \|\alpha \mathbf{h} - \mathbf{a}\|^2$$

## Compute-and-Forward: Fading Channels – Illustration

Compute sum of lattice points modulo the coarse lattice:



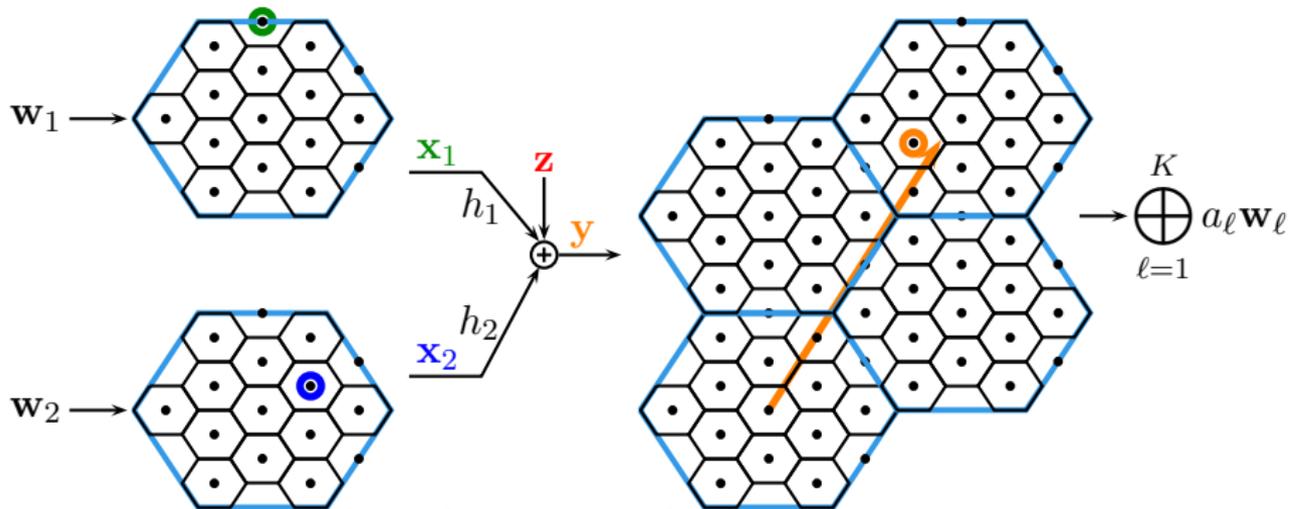
$$\alpha \mathbf{h} = [ \alpha 1.4 \quad \alpha 2.1 ]$$

$$\mathbf{a} = [ 2 \quad 3 ]$$

$$\text{Effective noise: } \alpha^2 N + P \|\alpha \mathbf{h} - \mathbf{a}\|^2$$

# Compute-and-Forward: Fading Channels – Illustration

Map back to equation of message symbols over the field:

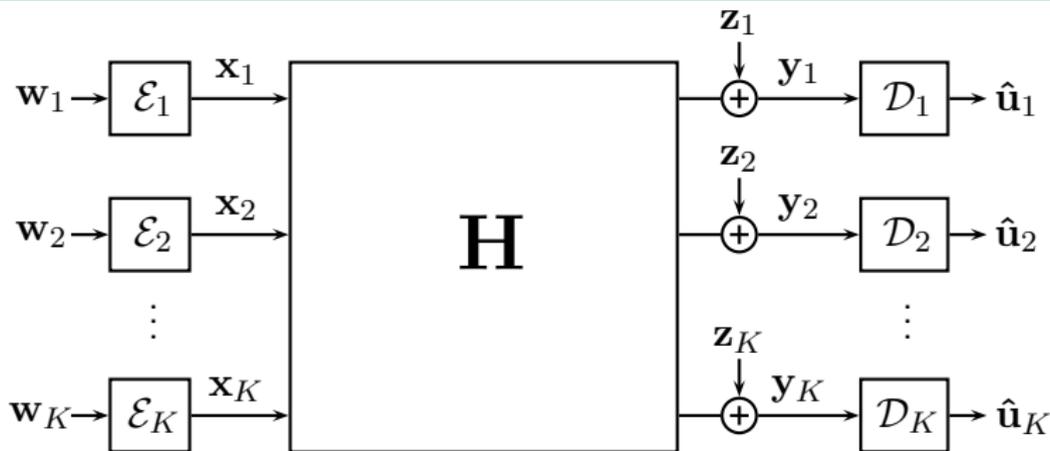


$$\alpha \mathbf{h} = [ \alpha_{1.4} \quad \alpha_{2.1} ]$$

$$\mathbf{a} = [ 2 \quad 3 ]$$

$$\text{Effective noise: } \alpha^2 N + P \|\alpha \mathbf{h} - \mathbf{a}\|^2$$

## Computation over Fading Channels – Multiple Receivers



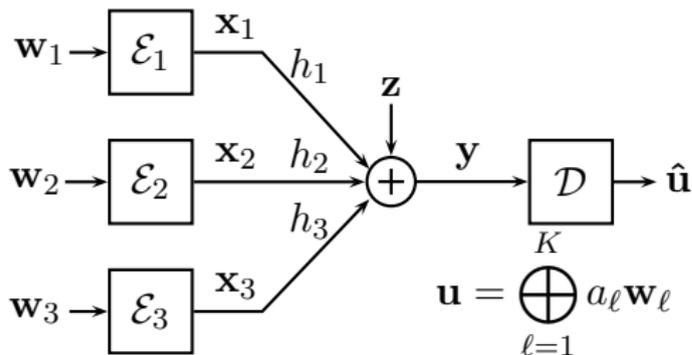
- Equal rates  $R$ . **No channel state information** (CSI) at transmitters.
- Receivers use their CSI to select coefficients, **decode linear equation**

$$\mathbf{u}_k = \bigoplus_{\ell=1}^K a_{k\ell} \mathbf{w}_\ell$$

- Reliable decoding possible if

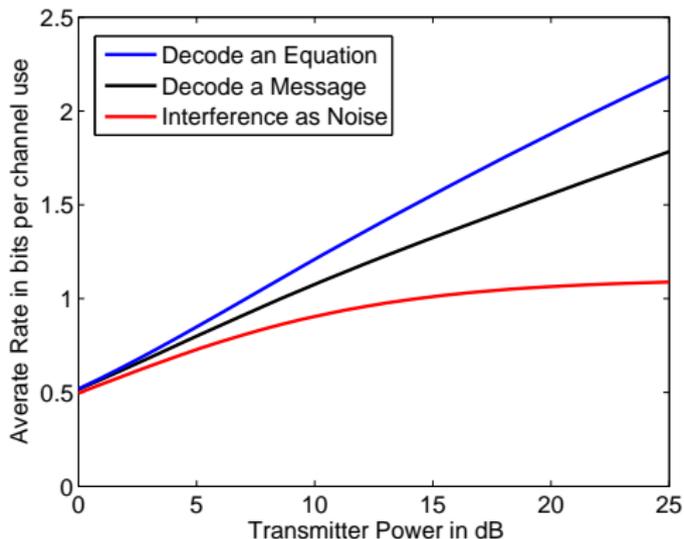
$$R < \min_{k: a_{k\ell} \neq 0} \frac{1}{2} \log \left( \frac{N + P \|\mathbf{h}_k\|^2}{N \|\mathbf{a}_k\|^2 + P(\|\mathbf{h}_k\|^2 \|\mathbf{a}_k\|^2 - (\mathbf{h}_k^T \mathbf{a}_k)^2)} \right)$$

# Computation over Fading Channels – No CSIT



Relay either decodes some **linear function of messages** or an **individual message**.

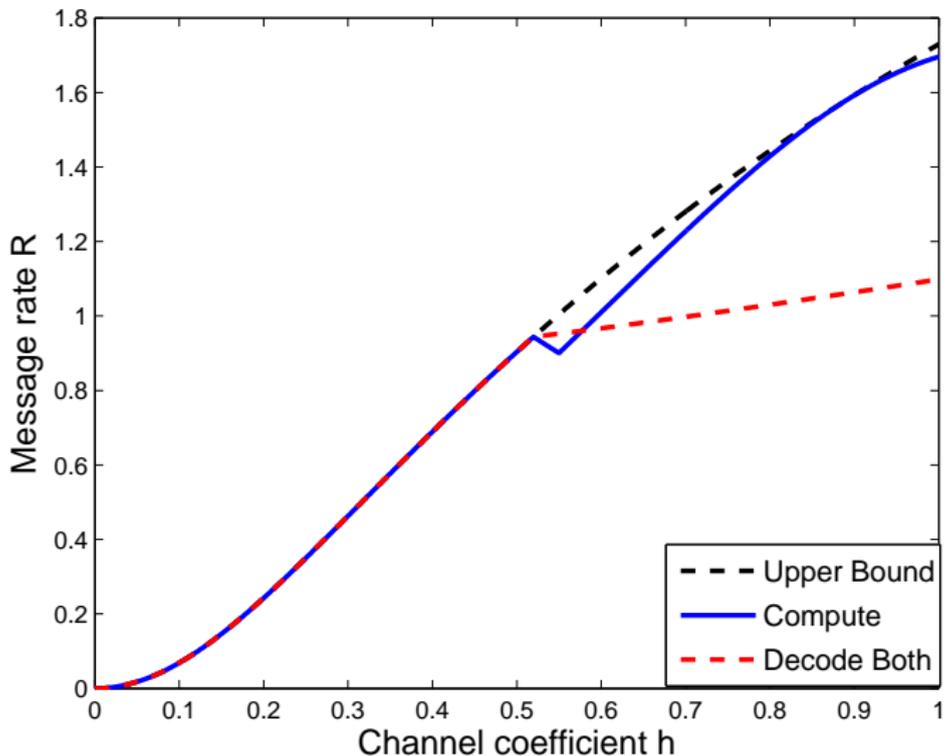
- Three transmitters that do not know the fading coefficients.
- Average rate plotted for i.i.d. Gaussian fading.



## Computation over Fading Channels – No CSIT

- Receiver observes  $\mathbf{y} = \mathbf{x}_1 + h\mathbf{x}_2 + \mathbf{z}$ .
- Recovers  $a\mathbf{w}_1 \oplus b\mathbf{w}_2$  for  $a, b \neq 0$ .

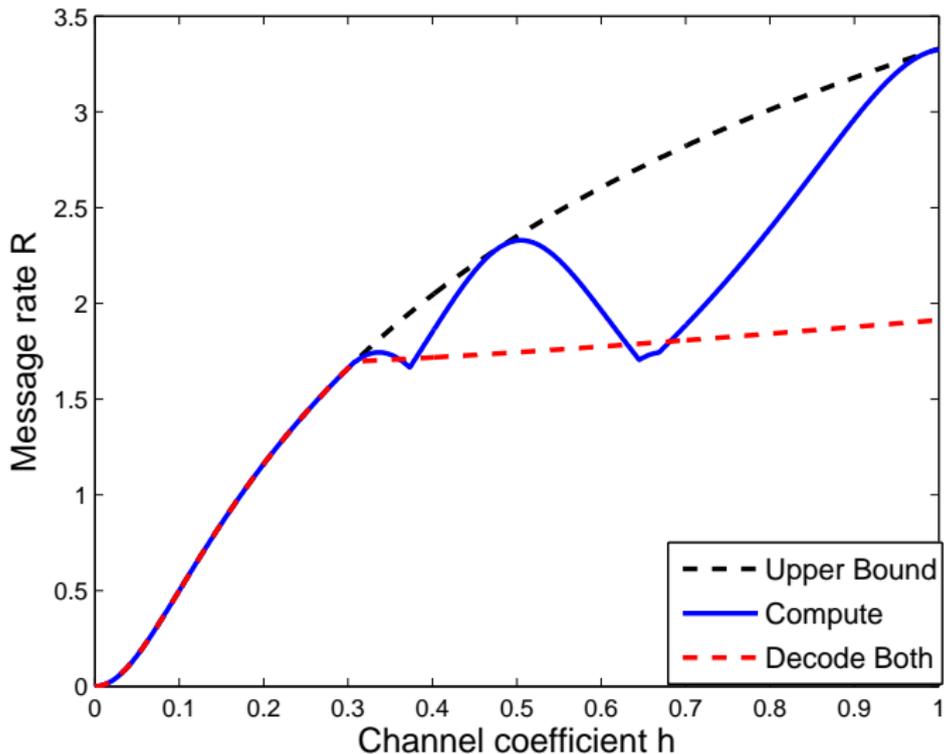
10dB



## Computation over Fading Channels – No CSIT

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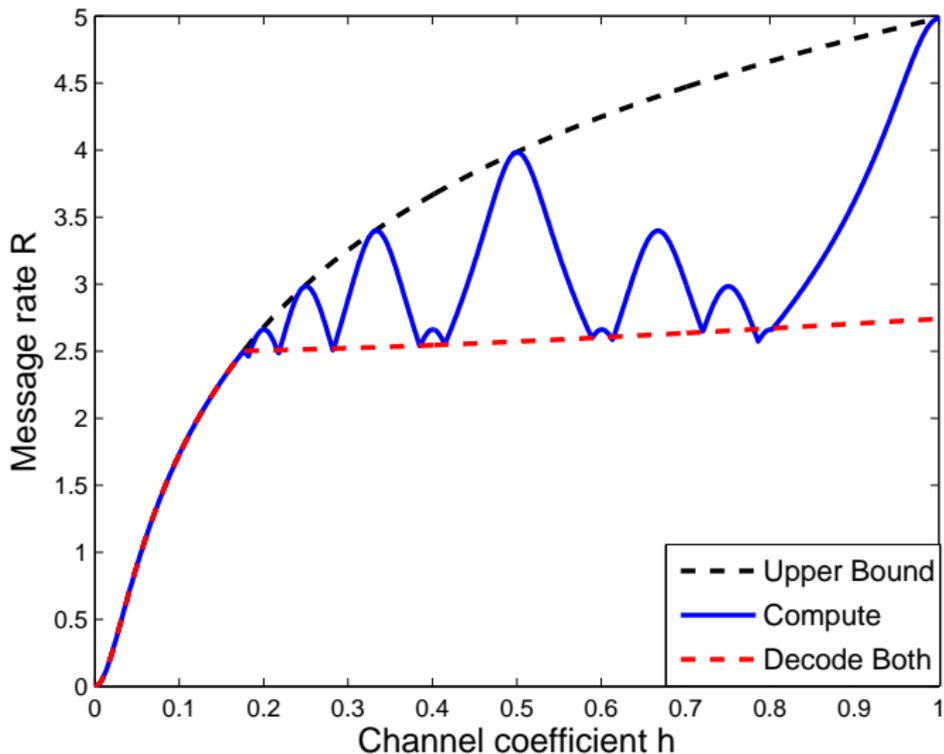
20dB



## Computation over Fading Channels – No CSIT

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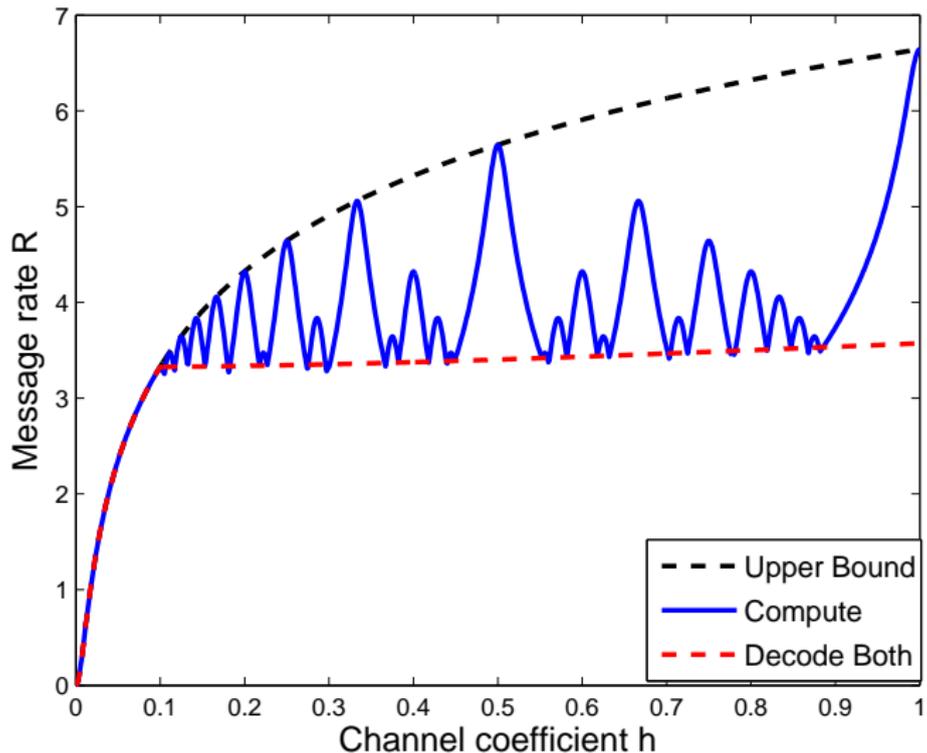
30dB



## Computation over Fading Channels – No CSIT

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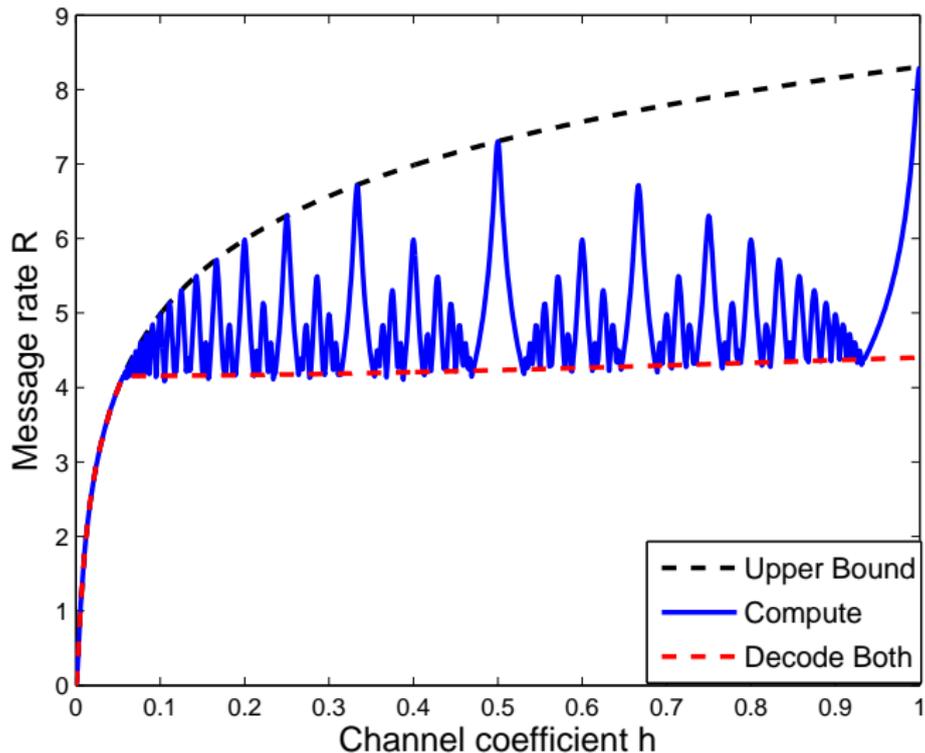
40dB



## Computation over Fading Channels – No CSIT

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- Recovers  $a\mathbf{w}_1 \oplus b\mathbf{w}_2$  for  $a, b \neq 0$ .

50dB



## *Can we fill in the valleys?*

- Compute-and-forward does well at **rational** coefficients and poorly at **irrational coefficients**.
- This is the opposite of the behavior observed in “real interference alignment” (**Motahari et al. '09**).
- As demonstrated in the previous talk, we can do better using superposition. This alters the effective channel gains.
- Next talk covers this issue in depth at high SNR.
- How about finite SNR? Similar issues as encountered in static interference channels.

## Successive Cancellation

- Receiver observes  $\mathbf{y} = \sum_{\ell=1}^L h_{\ell} \mathbf{x}_{\ell} + \mathbf{z}$

Successive cancellation:

- Decode  $\mathbf{x}_i$ .
- Calculate  $\mathbf{y} - h_i \mathbf{x}_i$ .
- Receiver now has

$$\sum_{\ell \neq i} h_{\ell} \mathbf{x}_{\ell} + \mathbf{z}$$

## Successive Cancellation Computation

- Receiver observes  $\mathbf{y} = \sum_{\ell=1}^L h_{\ell} \mathbf{x}_{\ell} + \mathbf{z}$

### Successive cancellation:

- Decode  $\mathbf{x}_i$ .
- Calculate  $\mathbf{y} - h_i \mathbf{x}_i$ .
- Receiver now has

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### Successive computation:

- Decode  $\sum_{\ell=1}^L a_{\ell} \mathbf{x}_{\ell}$ .
- Calculate  $\mathbf{y} + \beta \sum_{\ell=1}^L a_{\ell} \mathbf{x}_{\ell}$ .
- Receiver now has

$$\sum_{\ell=1}^L (h_{\ell} + \beta a_{\ell}) \mathbf{x}_{\ell} + \mathbf{z}$$

## Successive Computation

- So far, we have only decoded a **modulo sum** of the lattice points:

$$\left[ \sum_{\ell} a_{\ell} \mathbf{t}_{\ell} \right] \bmod \Lambda .$$

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$$\left[ \left[ \sum_{\ell} a_{\ell} \mathbf{t}_{\ell} \right] \bmod \Lambda + \left[ \sum_{\ell} a_{\ell} \mathbf{d}_{\ell} \right] \bmod \Lambda \right] \bmod \Lambda = \left[ \sum_{\ell} a_{\ell} \mathbf{x}_{\ell} \right] \bmod \Lambda$$

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- Subtract this from  $\mathbf{y}$  to expose the **coarse lattice point** nearest to the **real sum**:

$$\mathbf{y} - \left[ \sum_{\ell} a_{\ell} \mathbf{x}_{\ell} \right] \bmod \Lambda = Q_{\Lambda} \left( \sum_{\ell} a_{\ell} \mathbf{x}_{\ell} \right) + \sum_{\ell} (h_{\ell} - a_{\ell}) \mathbf{x}_{\ell} + \mathbf{z}$$

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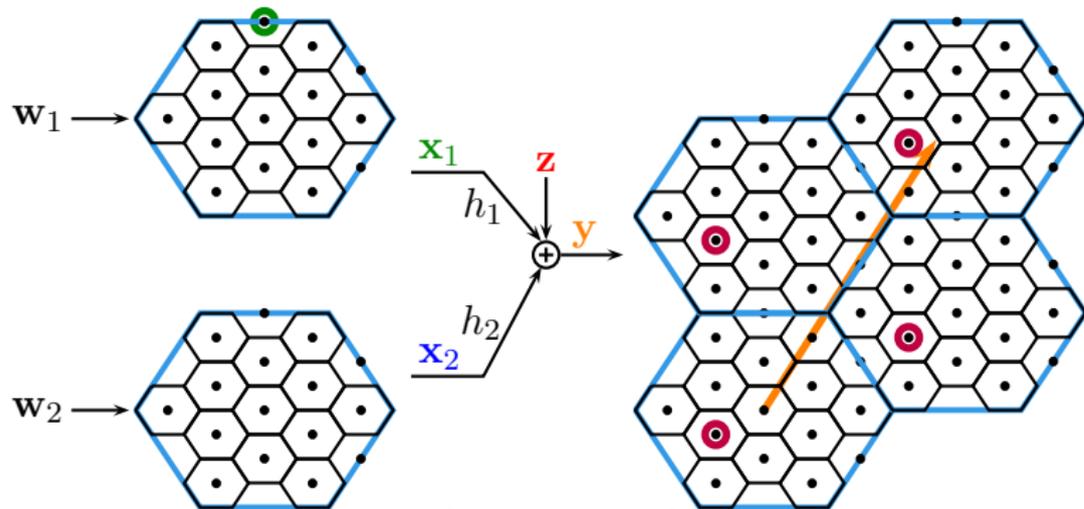
$$\mathbf{y} - \left[ \sum_{\ell} a_{\ell} \mathbf{x}_{\ell} \right] \bmod \Lambda = Q_{\Lambda} \left( \sum_{\ell} a_{\ell} \mathbf{x}_{\ell} \right) + \sum_{\ell} (h_{\ell} - a_{\ell}) \mathbf{x}_{\ell} + \mathbf{z}$$

- Coarse lattice point easier to decode than fine lattice point:

$$Q_{\Lambda} \left( Q_{\Lambda} \left( \sum_{\ell} a_{\ell} \mathbf{x}_{\ell} \right) + \sum_{\ell} (h_{\ell} - a_{\ell}) \mathbf{x}_{\ell} + \mathbf{z} \right) = Q_{\Lambda} \left( \sum_{\ell} a_{\ell} \mathbf{x}_{\ell} \right) \quad \text{w.h.p.}$$

## Successive Computation Illustration

We have the **modulo sum**.



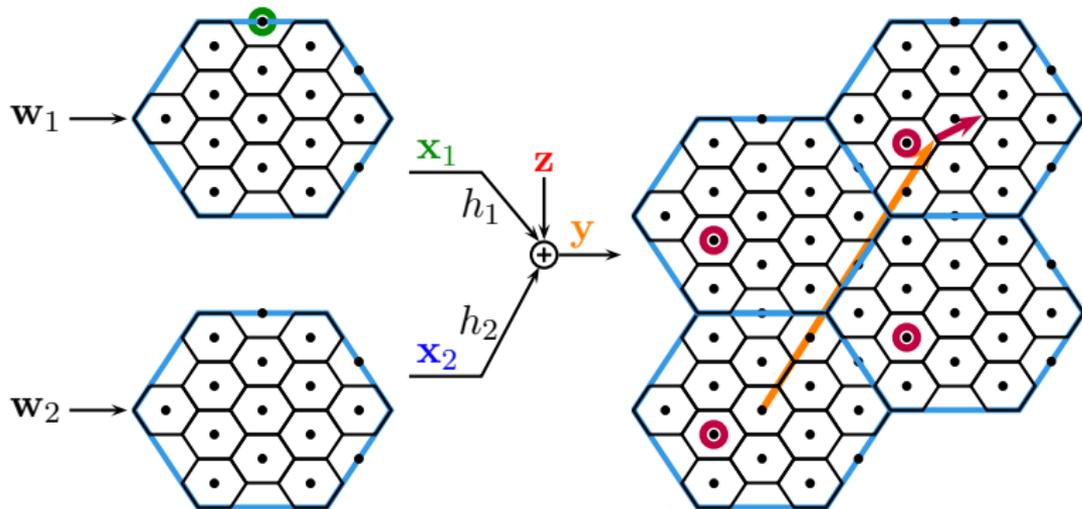
$$\alpha \mathbf{h} = [ \alpha 1.4 \quad \alpha 2.1 ]$$

$$\mathbf{a} = [ 2 \quad 3 ]$$

$$\text{Effective noise: } \alpha^2 N + P \|\alpha \mathbf{h} - \mathbf{a}\|^2$$

## Successive Computation Illustration

Subtract **modulo sum** from the received signal.



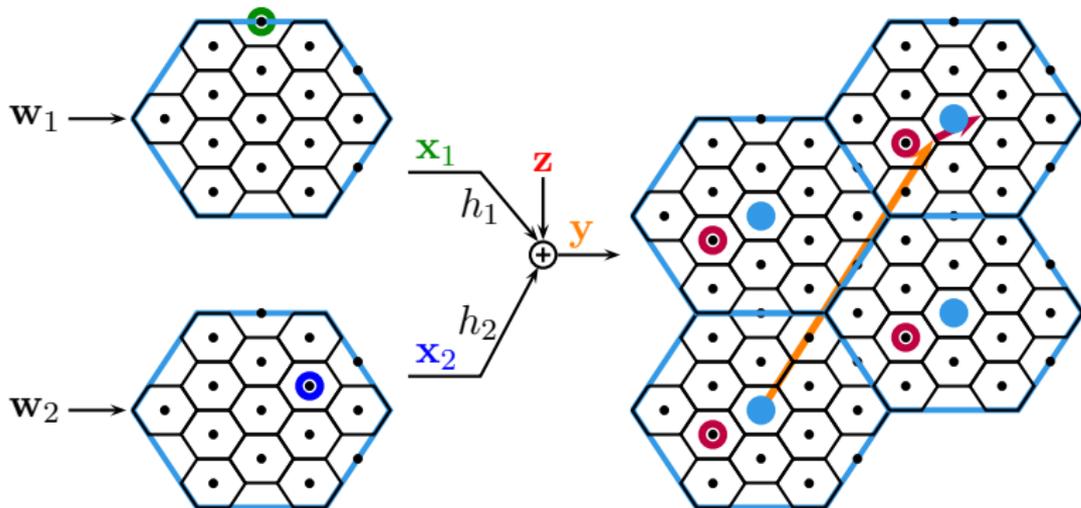
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## Successive Computation Illustration

Decode to the closest **coarse lattice point**.



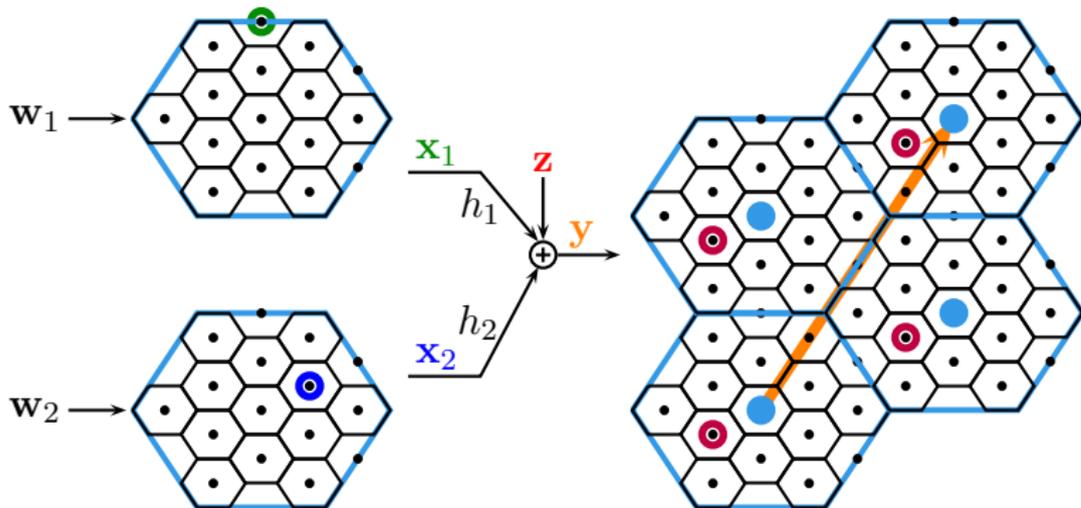
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## Successive Computation Illustration

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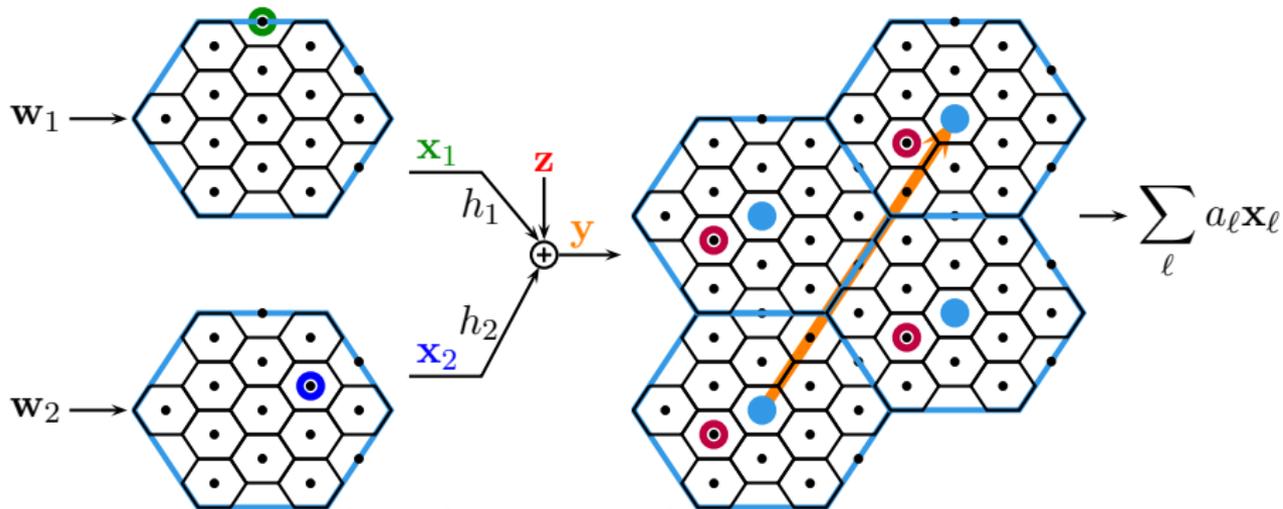
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## Successive Computation Illustration

Now we can infer the real sum.



$$\alpha \mathbf{h} = [ \alpha 1.4 \quad \alpha 2.1 ]$$

$$\mathbf{a} = [ 2 \quad 3 ]$$

$$\text{Effective noise: } \alpha^2 N + P \|\alpha \mathbf{h} - \mathbf{a}\|^2$$

- Finally, we get back the real sum:

$$\left[ \sum_{\ell} a_{\ell} \mathbf{x}_{\ell} \right] \bmod \Lambda + Q_{\Lambda} \left( \sum_{\ell} a_{\ell} \mathbf{x}_{\ell} \right) = \sum_{\ell} a_{\ell} \mathbf{x}_{\ell}$$

- What can we do with this?
- Change the effective channel gains and decode a new equation.
- Recover some existing results on interference alignment.

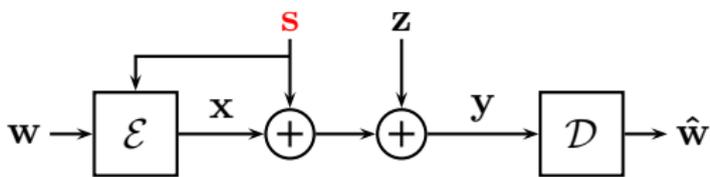
## Alignment via Successive Computation

- Assume each receiver observes  $\mathbf{y}_k = h\mathbf{x}_k + \sum_{\ell \neq k} \mathbf{x}_\ell + \mathbf{z}_k$ .
- Decode equation of the form  $p\mathbf{x}_k + \sum_{\ell \neq k} q\mathbf{x}_\ell$ .
- Rate  $\frac{1}{2} \log \left( \frac{\text{SNR}}{q^2 + \text{SNR}|qh - p|^2} \right) \leq \frac{1}{2} \log \left( \frac{\text{SNR}}{q^2 + \text{SNR}/q^2} \right)$
- Plug in  $q \approx \text{SNR}^{1/4}$  to get  $R \approx \frac{1}{4} \log(P)$ .
- Calculate  $\mathbf{y}_k - \frac{p}{q}\mathbf{x}_k + \sum_{\ell \neq k} \mathbf{x}_\ell = \left( h - \frac{p}{q} \right) \mathbf{x}_k + \mathbf{z}_k$
- By Khinchin's Theorem, residual channel coefficient allows  $R \approx \frac{1}{4} \log(P)$ .

## Dirty Paper Coding

$\mathbf{s}$  is **interference** known noncausally to the encoder.

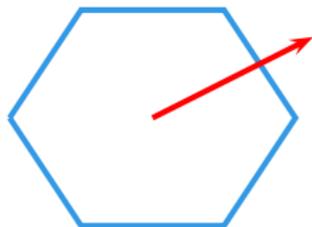
Assume  $\mathbf{s}$  i.i.d. Gaussian, very large variance  $P_S$ .



**Erez-Shamai-Zamir '05:**

Encoder subtracts  $\alpha\mathbf{s}$ , dithers, and takes  $\bmod \Lambda$ .

$$\mathbf{x} = [\mathbf{t} - \alpha\mathbf{s} + \mathbf{d}] \bmod \Lambda$$



Decoder scales by  $\alpha$ , removes dither, takes  $\bmod \Lambda$ , and recovers  $\mathbf{t}$ . **Interference is cancelled.**

$$\begin{aligned} [\alpha\mathbf{y} - \mathbf{d}] \bmod \Lambda &= [\mathbf{x} + \alpha\mathbf{s} - \mathbf{d} + \mathbf{z} - (1 - \alpha)\mathbf{x}] \bmod \Lambda \\ &= \left[ [\mathbf{t} - \alpha\mathbf{s} + \mathbf{d}] \bmod \Lambda + \alpha\mathbf{s} - \mathbf{d} + \mathbf{z} - (1 - \alpha)\mathbf{x} \right] \bmod \Lambda \\ &= \left[ \mathbf{t} + \mathbf{z} - (1 - \alpha)\mathbf{x} \right] \bmod \Lambda \end{aligned}$$

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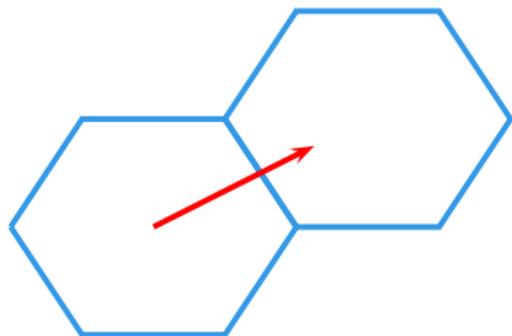
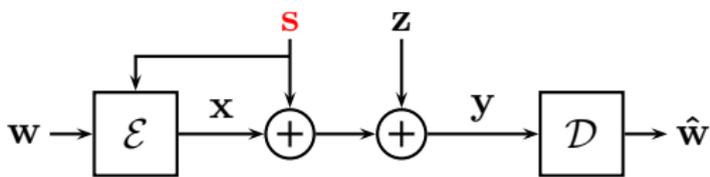
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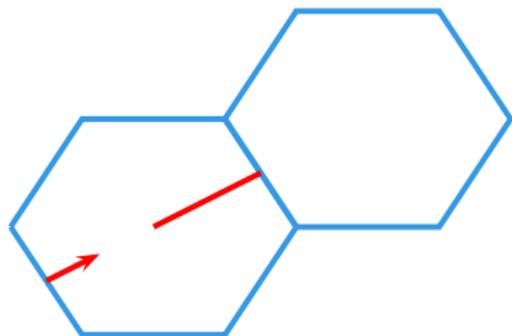
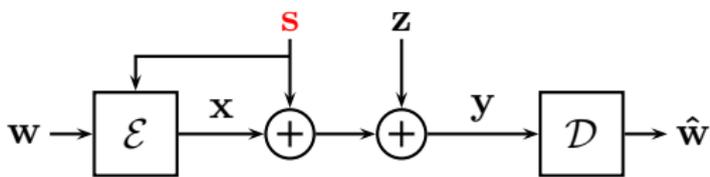
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Decoder scales by  $\alpha$ , removes dither, takes  $\bmod \Lambda$ , and recovers  $\mathbf{t}$ .  
**Interference is cancelled.**

$$\begin{aligned} [\alpha\mathbf{y} - \mathbf{d}] \bmod \Lambda &= [\mathbf{x} + \alpha\mathbf{s} - \mathbf{d} + \mathbf{z} - (1 - \alpha)\mathbf{x}] \bmod \Lambda \\ &= \left[ [\mathbf{t} - \alpha\mathbf{s} + \mathbf{d}] \bmod \Lambda + \alpha\mathbf{s} - \mathbf{d} + \mathbf{z} - (1 - \alpha)\mathbf{x} \right] \bmod \Lambda \\ &= \left[ \mathbf{t} + \mathbf{z} - (1 - \alpha)\mathbf{x} \right] \bmod \Lambda \end{aligned}$$



## Dirty Paper Coding

$\mathbf{s}$  is **interference** known noncausally to the encoder.

Assume  $\mathbf{s}$  i.i.d. Gaussian, very large variance  $P_S$ .

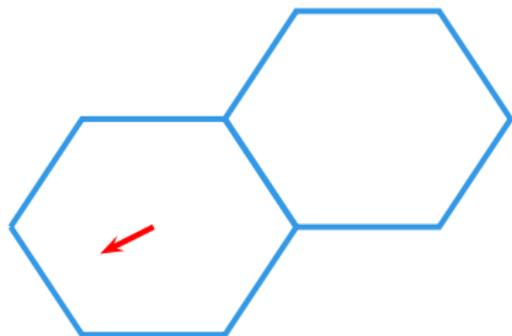
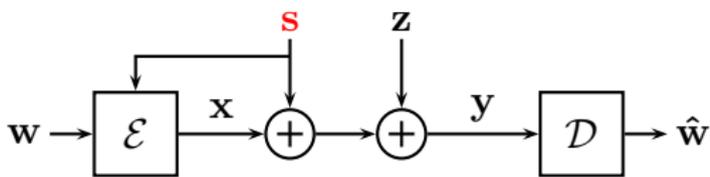
**Erez-Shamai-Zamir '05:**

Encoder subtracts  $\alpha\mathbf{s}$ , dithers, and takes  $\bmod \Lambda$ .

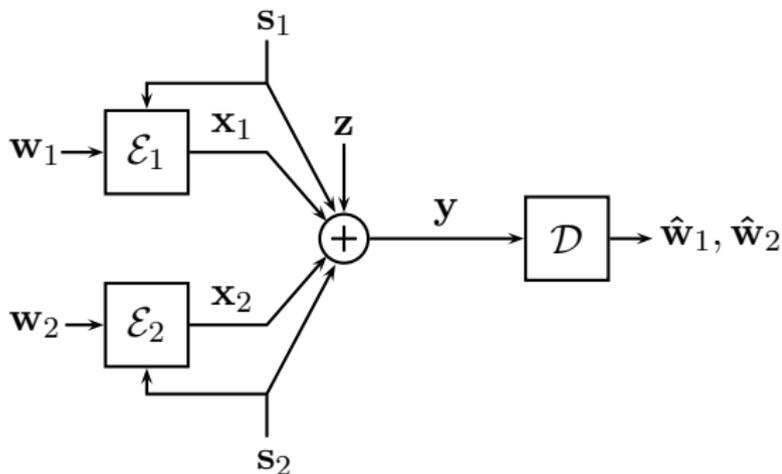
$$\mathbf{x} = [\mathbf{t} - \alpha\mathbf{s} + \mathbf{d}] \bmod \Lambda$$

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**Interference is cancelled.**

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## Dirty Gaussian Multiple-Access Channel



### Philosof-Zamir-Erez-Khisti '11:

- Encoder 1 knows interference  $s_1$ .
- Encoder 2 knows interference  $s_2$ .
- Need to **cancel out interference** in a **distributed** fashion.
- Assume i.i.d. Gaussian interference with very large variance  $P_S$ . Random i.i.d. methods yield rate that goes to 0 as  $P_S$  goes to infinity.

## Dirty Gaussian Multiple-Access Channel

Subtract (part of) the **interference signals** ahead of time:

$$\mathbf{x}_1 = [\mathbf{t}_1 - \alpha \mathbf{s}_1 + \mathbf{d}_1] \bmod \Lambda$$

$$\mathbf{x}_2 = [\mathbf{t}_2 - \alpha \mathbf{s}_2 + \mathbf{d}_2] \bmod \Lambda$$

Decoder removes dithers:

$$[\alpha \mathbf{y} - \mathbf{d}_1 - \mathbf{d}_2] \bmod \Lambda$$

$$= [\alpha(\mathbf{x}_1 + \mathbf{x}_2 + \mathbf{s}_1 + \mathbf{s}_2 + \mathbf{z}) - \mathbf{d}_1 - \mathbf{d}_2] \bmod \Lambda$$

$$= [\mathbf{x}_1 + \mathbf{x}_2 + \alpha(\mathbf{s}_1 + \mathbf{s}_2) - (1 - \alpha)(\mathbf{x}_1 + \mathbf{x}_2) + \alpha \mathbf{z}] - \mathbf{d}_1 - \mathbf{d}_2] \bmod \Lambda$$

$$= [\mathbf{t}_1 + \mathbf{t}_2 + (1 - \alpha)(\mathbf{x}_1 + \mathbf{x}_2) + \alpha \mathbf{z}] \bmod \Lambda$$

Select  $\alpha = 2P/(2P + N)$  to obtain

$$R_1 + R_2 \leq \frac{1}{2} \log \left( \frac{1}{2} + \frac{P}{N} \right)$$

**Maple Syrup Problem:** Prove this is the best possible

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**Maple Syrup Problem:** Prove this is the best possible (or find a better achievable scheme).

- Superposition (**Shlomo's Talk**)
- Fading
- Successive Cancellation
- Dirty Paper Coding
- Joint Decoding (**Uri's Talk**)
- List Decoding (**Natasha's Talk**)

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- Outer bounds?