

Augmented Lattice Reduction for MIMO decoding

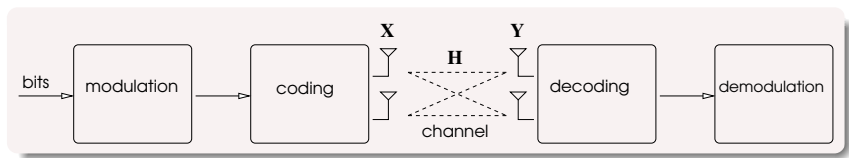
LAURA LUZZI

joint work with G. Rekaya-Ben Othman and J.-C. Belfiore
at Télécom-ParisTech

NANYANG TECHNOLOGICAL UNIVERSITY
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- Multiple antenna systems allow to improve data rates and reliability.
- In order to increase data rates, both the number of antennas and the size of the signal set can be increased.
- This entails a high **decoding complexity** which is a real challenge for practical implementation.

- 1 MIMO systems
- 2 Decoding
 - Lattice decoding
 - Lattice reduction-aided decoding
- 3 Augmented lattice reduction
 - Method
 - Performance
 - Complexity
- 4 Open problems



- **multiplexing gain**

- send independent data on each antenna
- improve the rate

- **diversity gain**

- send independent copies of the same data
- improve reliability

System model: spatial multiplexing

$$\mathbf{y}_{n \times 1} = \mathbf{H}_{n \times m} \mathbf{x}_{m \times 1} + \mathbf{w}_{n \times 1}$$

received signal channel codeword noise

- m transmit antennas, n receive antennas
- \mathbf{H} , \mathbf{w} are random matrices with i.i.d. complex Gaussian entries
- $x_i \in \mathcal{S} \subset \mathbb{Z}[i]$ is the signal transmitted by antenna i

$$\mathbf{Y}_{n \times t} = \mathbf{H}_{n \times m} \mathbf{X}_{m \times t} + \mathbf{W}_{n \times t}$$

received signal channel codeword noise

- m transmit antennas, n receive antennas, t frame length
- codewords are represented by matrices or **space-time blocks**
- the matrix element $x_{i,j} \in \mathbb{C}$ represents the signal sent by antenna i at time j

$$d = - \lim_{\text{SNR} \rightarrow \infty} \frac{\log(P_e)}{\log(\text{SNR})},$$

where P_e is the error probability, and SNR is the signal-to-noise ratio.

Maximum diversity order

- For spatial multiplexing:

$$d_{\max} = n \quad (\text{receive diversity})$$

- For space-time coding:

$$d_{\max} = mn \quad (\text{transmit and receive diversity})$$

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Spatial multiplexing case

$$\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{w}$$

The received vector is the translated version of a point in the lattice generated by \mathbf{H} .

Coded case

$$\mathbf{Y} = \mathbf{H}\mathbf{X} + \mathbf{W}$$

Equivalent lattice formulation after vectorizing matrices:

$$\mathbf{y} = \mathbf{H}_l\Phi\mathbf{s} + \mathbf{w}$$

- \mathbf{H}_l linear map corresponding to left multiplication by \mathbf{H}
- Φ generator matrix of the code
- \mathbf{s} vector of information signals

Optimal decoding amounts to solving the **Closest Vector Problem (CVP)** in the lattice generated by \mathbf{H} :

$$\hat{\mathbf{x}} = \underset{\mathbf{x}' \in \mathcal{S}}{\operatorname{argmin}} \|\mathbf{y} - \mathbf{H}\mathbf{x}'\|^2 \quad \text{ML solution}$$

- **ML decoders**
 - Sphere Decoder, Schnorr-Euchner algorithm...
 - optimal performance but exponential complexity
- **Suboptimal decoders**
 - zero forcing (ZF), successive interference cancellation (SIC)...
 - polynomial complexity, but they don't attain maximal diversity

Example: ZF decoding

$$\mathbf{y} = \mathbf{H}\mathbf{x} + \mathbf{w}$$

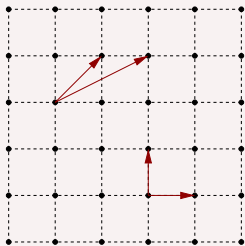
$$\hat{\mathbf{x}}_{\text{ZF}} = \lfloor \mathbf{H}^{-1}\mathbf{y} \rfloor = \lfloor \mathbf{x} + \mathbf{H}^{-1}\mathbf{w} \rfloor$$

- if \mathbf{H} is orthogonal, ZF decoding is optimal
- if \mathbf{H} is ill-conditioned, the noise $\mathbf{H}^{-1}\mathbf{w}$ is amplified
- **Solution:** channel preprocessing by lattice reduction improves the performance of suboptimal decoders

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Lattice reduction

Two matrices \mathbf{H} and \mathbf{H}' generate the same lattice if $\mathbf{H}' = \mathbf{HT}$ with \mathbf{T} unimodular.



Lattice reduction:
find a lattice basis formed by
“short” and “nearly orthogonal”
vectors

- the most popular lattice reduction algorithm is the **LLL algorithm**, thanks to its polynomial complexity

Gram-Schmidt Orthogonalization (GSO)

```
h1* ← h1  
for  $i = 2, \dots, m$  do  
  for  $j = 1, \dots, i - 1$  do  
     $\mu_{i,j} \leftarrow \frac{\langle \mathbf{h}_i, \mathbf{h}_j^* \rangle}{\|\mathbf{h}_j^*\|^2}$   
  end  
   $\mathbf{h}_i^* \leftarrow \mathbf{h}_i - \sum_{j=1}^{i-1} \mu_{i,j} \mathbf{h}_j^*$   
end
```

LLL-reduced basis

H is LLL-reduced if its GSO satisfies the following properties:

- **Size reduction:** $|\mu_{k,l}| \leq \frac{1}{2}, \quad 1 \leq l < k \leq m,$
- **Lovasz condition:** $\|\mathbf{h}_k^* + \mu_{k,k-1} \mathbf{h}_{k-1}^*\|^2 \geq \frac{3}{4} \|\mathbf{h}_{k-1}^*\|^2, \quad 1 < k \leq m$

The LLL algorithm

Compute GSO

$k \leftarrow 2$

while $k \leq m$ **do**

 Size reduction RED($k, k-1$)

if *Lovasz($k, k-1$) is not satisfied* **then**

 swap \mathbf{h}_k and \mathbf{h}_{k-1}

 update GSO

$k \leftarrow \max(k - 1, 2)$

end

else

for $l = k - 2, \dots, 1$ **do**

 Size reduction RED(k, l)

end

$k \leftarrow k + 1$

end

end

Preprocessing using LLL reduction

m transmit antennas, n receive antennas

$$\mathbf{y}_{n \times 1} = \mathbf{H}_{n \times m} \mathbf{x}_{m \times 1} + \mathbf{w}_{n \times 1}$$

received signal channel codeword noise

$$\mathbf{H}_{\text{red}} = \mathbf{H}\mathbf{U} \quad \text{LLL-reduced form}$$

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LLL-ZF decoder

- compute the pseudo-inverse $\mathbf{H}_{\text{red}}^\dagger$
- $\hat{\mathbf{x}}_{\text{LLL-ZF}} = \mathbf{U} \left(\left[\mathbf{H}_{\text{red}}^\dagger \mathbf{y} \right] \right)$

Preprocessing using LLL reduction

m transmit antennas, n receive antennas

$$\begin{array}{ccccccc} \mathbf{y}_{n \times 1} & = & \mathbf{H}_{n \times m} & \mathbf{x}_{m \times 1} & + & \mathbf{w}_{n \times 1} \\ \text{received signal} & & \text{channel} & \text{codeword} & & \text{noise} \end{array}$$

$$\mathbf{H}_{\text{red}} = \mathbf{H}\mathbf{U} \quad \text{LLL-reduced form}$$

LLL-ZF decoder

- compute the pseudo-inverse $\mathbf{H}_{\text{red}}^\dagger$
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LLL-SIC decoder

- QR decomposition of \mathbf{H}_{red}
- $\tilde{\mathbf{y}} = \mathbf{Q}^H \mathbf{y} = \mathbf{R}\mathbf{x} + \mathbf{Q}^H \mathbf{w}$
- recursively compute $\tilde{x}_m = \left\lfloor \frac{\tilde{y}_m}{r_{mm}} \right\rfloor$,
 $\tilde{x}_i = \left\lfloor \frac{\tilde{y}_i - \sum_{j=i+1}^m r_{ij} \tilde{x}_j}{r_{ii}} \right\rfloor, i = m-1, \dots, 1$
- $\hat{\mathbf{x}}_{\text{LLL-SIC}} = \mathbf{U}\tilde{\mathbf{x}}$

Proposition [Taherzadeh, Mobasher, Khandani 2007]

LLL-ZF and LLL-SIC decoders attain the maximal receive diversity n .

- the average complexity of LLL-aided decoding is polynomial, which makes it an attractive technique
- however, LLL becomes less effective for high-dimensional lattices
- as a consequence, the performance gap with respect to ML decoding increases greatly when the number of antennas is large

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- new technique which combines preprocessing and detection: lattice reduction is used directly to decode
- MIMO decoding amounts to solving the **closest vector problem (CVP)** in the lattice generated by the channel matrix
- reduce the CVP to the SVP (**shortest vector problem**)
- use **LLL reduction** to solve the SVP in polynomial time

Embedding method to reduce the CVP to the SVP

- Follow Kannan's approach (1987): **embed** the m -dimensional lattice generated by \mathbf{H} into a suitable $(m + 1)$ -dimensional lattice

$$\tilde{\mathbf{H}} = \begin{pmatrix} \mathbf{H} & -\mathbf{y} \\ \mathbf{0} & t \end{pmatrix} \quad \text{augmented matrix}$$

- $\mathbf{v} = \begin{pmatrix} \mathbf{H}\mathbf{x} - \mathbf{y} \\ t \end{pmatrix} = \begin{pmatrix} \mathbf{w} \\ t \end{pmatrix} = \tilde{\mathbf{H}} \begin{pmatrix} \mathbf{x} \\ 1 \end{pmatrix}$
- **Strategy:** if $\|\mathbf{w}\|$ and t are “very small”, \mathbf{v} is the shortest vector in the augmented lattice. By finding \mathbf{v} , we recover \mathbf{x} .

Using LLL reduction to find the shortest lattice vector

- LLL-reduce $\tilde{\mathbf{H}}$: $\tilde{\mathbf{H}}_{\text{red}} = \tilde{\mathbf{H}}\tilde{\mathbf{U}}$

- **Problem:** in general, there's no guarantee that the shortest vector belongs to the reduced basis. However, the first column of $\tilde{\mathbf{H}}_{\text{red}}$ satisfies

$$\left\| \tilde{\mathbf{h}}_1^{\text{red}} \right\| \leq 2^{\frac{m}{2}} d_{\tilde{\mathbf{H}}}$$

- \mathbf{v} shortest vector in \mathcal{L} is called α -unique if $\forall \mathbf{u} \in \mathcal{L}$,

$$\|\mathbf{u}\| \leq \alpha \|\mathbf{v}\| \Rightarrow \mathbf{u}, \mathbf{v} \text{ linearly dependent}$$

Exponential gap technique

\mathbf{v} is $2^{\frac{m}{2}}$ -unique $\Rightarrow \pm \mathbf{v}$ is the first column of $\tilde{\mathbf{H}}_{\text{red}}$

LLL reduction finds the shortest lattice vector

$$a(\mathbf{H}) \doteq \min_{1 \leq i \leq m} \|\mathbf{h}_i^*\| \quad \Rightarrow \quad \frac{d_{\mathbf{H}}}{2^{\frac{m-1}{2}}} \leq a(\mathbf{H}_{\text{red}}) \leq d_{\mathbf{H}}$$

Lemma

Let $t = \frac{a(\mathbf{H}_{\text{red}})}{2^{m+1}}$, and suppose that $\|\mathbf{w}\| \leq \frac{d_{\mathbf{H}}}{2^{m+1}}$.

Then $\mathbf{v} = \begin{pmatrix} \mathbf{H}\mathbf{x} - \mathbf{y} \\ t \end{pmatrix}$ is a $2^{\frac{m}{2}}$ -unique shortest vector in $\mathcal{L}(\tilde{\mathbf{H}})$.

Sketch of the proof:

Suppose $\exists \mathbf{u} = \begin{pmatrix} \mathbf{H}\mathbf{x}' - q\mathbf{y} \\ qt \end{pmatrix}$ such that $\|\mathbf{u}\| \leq 2^{\frac{m}{2}} \|\mathbf{v}\|$, and \mathbf{u}, \mathbf{v} linearly independent

- $\|\mathbf{u}\| \geq |q|t \Rightarrow |q| \leq \frac{2^{\frac{m}{2}} \|\mathbf{v}\|}{t}$
- $\|\mathbf{H}(\mathbf{x}' - q\mathbf{x})\| \leq \|\mathbf{H}\mathbf{x}' - q\mathbf{y}\| + |q| \|\mathbf{y} - \mathbf{H}\mathbf{x}\| \leq 2^{\frac{m}{2}} \|\mathbf{v}\| + \frac{2^{\frac{m}{2}} \|\mathbf{v}\|}{t} \|\mathbf{w}\| \leq 2^{\frac{m}{2}} \sqrt{\|\mathbf{w}\|^2 + t^2} \left(1 + \frac{\|\mathbf{w}\|}{t}\right) < d_{\mathbf{H}} \Rightarrow \text{contradiction.}$

- Choose $t = \frac{a(\mathbf{H}_{\text{red}})}{2^{m+1}} \Rightarrow$ the augmented lattice has an exponential gap
- LLL-reduce $\tilde{\mathbf{H}}$: $\tilde{\mathbf{H}}_{\text{red}} = \tilde{\mathbf{H}}\tilde{\mathbf{U}}$
- If $\|\mathbf{w}\| \leq \frac{d_{\mathbf{H}}}{2^{m+1}}$, $\mathbf{v} = \tilde{\mathbf{H}} \begin{pmatrix} \mathbf{x} \\ 1 \end{pmatrix}$ is the first column of $\tilde{\mathbf{H}}_{\text{red}}$
- find the transmitted message \mathbf{x} on the first column of $\tilde{\mathbf{U}}$:

$$\hat{\mathbf{x}} = \frac{1}{\tilde{u}_{m+1,1}} (\tilde{u}_{1,1}, \dots, \tilde{u}_{m,1})^T$$

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Diversity gain

$$d = - \lim_{\text{SNR} \rightarrow \infty} \frac{\log(P_e)}{\log(\text{SNR})}$$

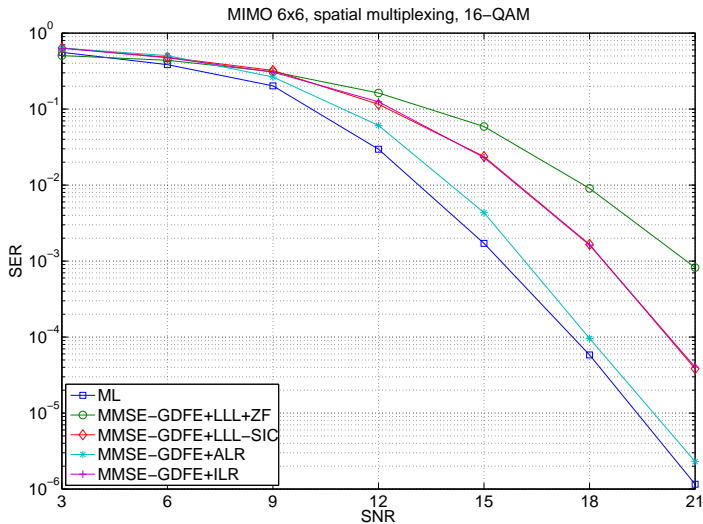
Proposition

If $t = \frac{a(\mathbf{H}_{\text{red}})}{2^{m+1}}$, then augmented lattice reduction achieves the **maximum receive diversity** n .

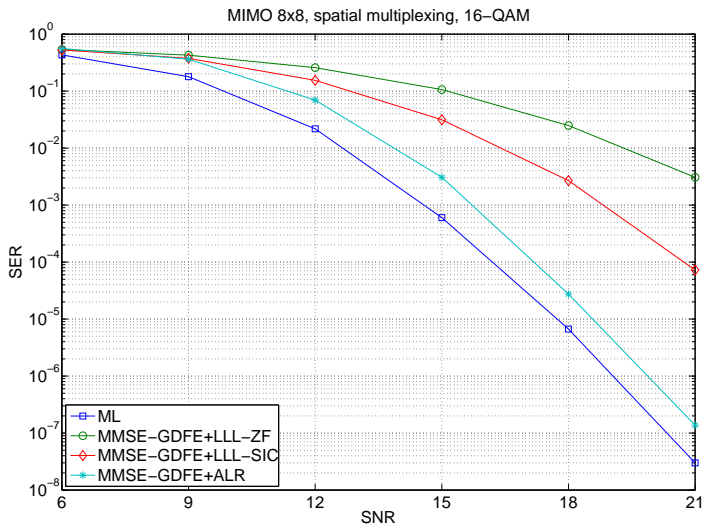
Sketch of the proof.

$$P_e(\mathbf{H}) \leq P \left\{ \|\mathbf{w}\| > \frac{d_{\mathbf{H}}}{2^{m+1}} \right\} \leq \frac{C(\log(\text{SNR}))^{n+1}}{\text{SNR}^n}$$

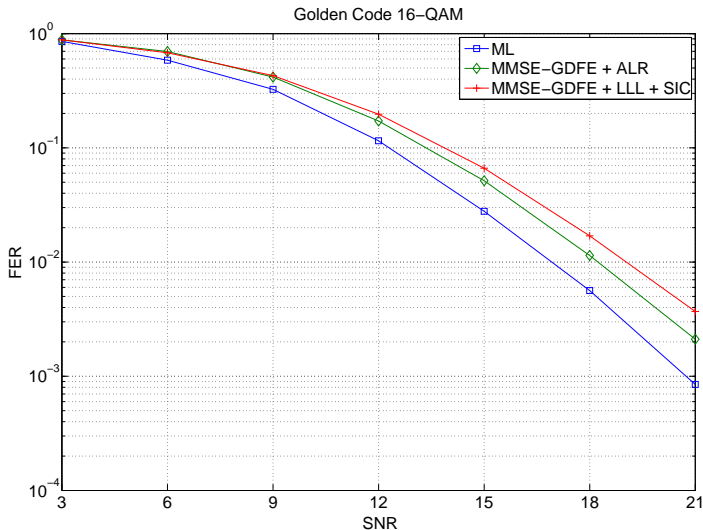
6×6 MIMO system, spatial multiplexing



8×8 MIMO system, spatial multiplexing



Coded case: 2×2 system, Golden Code



[N. Kim, H. Park, "Improved lattice reduction aided detections for MIMO systems", *Vehicular Technology Conference* 2006]

- **Same form** of the augmented matrix $\tilde{\mathbf{H}}$
- **No exponential gap technique:**
 - the parameter t is "big" ($t > \max |r_{ii}|$, where $\mathbf{H}_{\text{red}} = \mathbf{QR}$)
 - the solution is found on the last column of $\tilde{\mathbf{U}}$
- no guarantee that LLL reduction will find the right vector. In fact, one can prove that the performance is the **same as LLL-SIC**.

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- average number of iterations of the LLL algorithm:

$$\mathbb{E}[K(\mathbf{H})] \sim O(n^2 \log n) \quad [\text{Jalden } et al. \text{ 2008}]$$

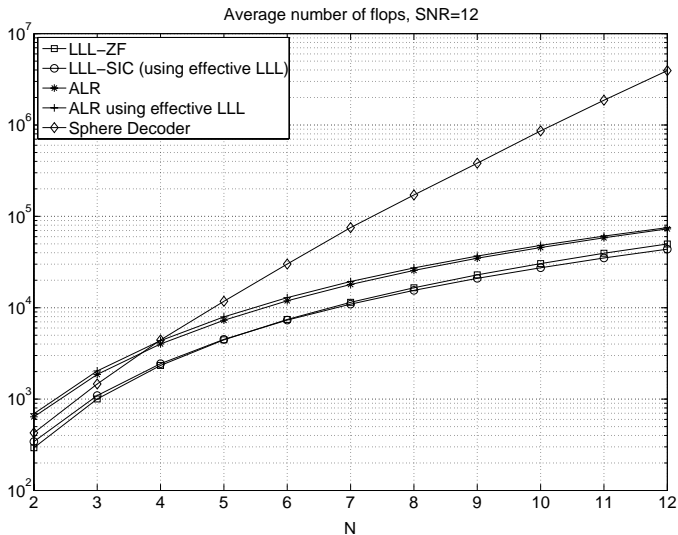
- each iteration requires $O(n^2)$ operations, which can be reduced to $O(n)$ for LLL-SIC [Ling, Howgrave-Graham 2007]

⇒ the total complexity of LLL-SIC is bounded by $O(n^3 \log n)$

Complexity bound for augmented lattice reduction

- $\mathbb{E}[K(\tilde{\mathbf{H}})] \leq O(n^3)$
- the total complexity is bounded by $O(n^4)$.

Complexity: numerical simulations



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- explain *why* the performance of augmented lattice reduction is better than LLL-SIC
- find more realistic bounds on the complexity
- for high-dimensional space-time codes, the gap between ML decoding and augmented lattice reduction is still huge.
How can we bridge this gap?

Thank you for listening!!