

## The Principle of Universal Lattice Decoding

presented by

Wai Ho Mow

Hong Kong University of Science & Technology

## Outline

- Preliminaries on lattices
- Closest Vector Problem (CVP)
- CVP in communications
- Lattice basis reduction
- Sphere decoding (SD)
- Lattice-reduced sub-optimal detectors
- Low-complexity MLD via packing radius test
- Simulation results for MIMO fading channels
- Concluding remarks

# Preliminary (1)

- For  $n \le m$ , let  $\mathbf{b}_1, \mathbf{b}_2, \dots, \mathbf{b}_n$  be a set of independent vectors in  $\mathbf{R}^m$
- **b**<sub>i</sub> are called the basis vectors
- A *lattice* is defined as the set of points:

$$\{\mathbf{x} \mid \mathbf{x} = a_1 \mathbf{b}_1 + \ldots + a_n \mathbf{b}_n\}$$

where  $a_i$  are integers

Equivalently, in matrix form:

where  $\mathbf{B} = [\mathbf{b}_1 \dots \mathbf{b}_n]$  and **a** is an integer (column) vector

The same lattice (i.e. the same set of points) can be generated by different basis:

 $L = L(\mathbf{B}_1) = L(\mathbf{B}_2)$  iff  $\mathbf{B}_1 = \mathbf{U}\mathbf{B}_2$ 

where **U** is an *unimodular* matrix (i.e.  $det(U) = \pm 1$ )

The determinant of lattice L is defined as the volume of the fundamental parallelotope of L:

$$det(L) = det(L(\mathbf{B})) = |det(\mathbf{B})|$$

# Preliminary (2)

- Gram-Schmidt Orthogonalization (GSO):
  - □ for any basis  $\mathbf{B} = [\mathbf{b}_1 \dots \mathbf{b}_n]$  we can find a set of orthogonal vectors  $\{\mathbf{b}_i^*\}$  that span the space of  $L(\mathbf{B})$ :

$$\mathbf{b}_{i}^{*} = \mathbf{b}_{i} - \sum_{j=i+1}^{n} \mu_{ij} \mathbf{b}_{j}^{*} \qquad \mu_{ij} = \frac{\langle \mathbf{b}_{i}, \mathbf{b}_{j}^{*} \rangle}{\left\| \mathbf{b}_{j}^{*} \right\|}$$
for *i* = *n*, *n*-1, ..., 1

- Note that:
  - □ different permutations of  $[\mathbf{b}_1 \dots \mathbf{b}_n]$  give you different set of  $\{\mathbf{b}^*_1 \dots \mathbf{b}^*_n\}$

$$\prod_{i=1}^{n} \left\| \mathbf{b}_{i}^{*} \right\| = \left| \det(\mathbf{B}) \right| = \det(L(\mathbf{B}))$$

## Closest Vector Problem (CVP)

#### Definition of the CVP:

□ Given a lattice *L*(**B**) and an arbitrary query point **q** in **R**<sup>*m*</sup>, to find, among all lattice points, the one that is closest to **q** w.r.t. Euclidean distance

□ More precisely, to solve:

$$\hat{\mathbf{x}} = \underset{\mathbf{x} \in L(\mathbf{B})}{\operatorname{arg\,min}} \|\mathbf{x} - \mathbf{q}\|^2$$

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# CVP in Communications (1)

- Many detection problems in communications can be reformulated as CVP.
- Detection for MIMO fading channels (Viterbo'93, Viterbo-Boutros'99, Damen et al.'2000):
  - assuming independent flat-fading channels, the received symbol vector y is given by:

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

where **x** := transmitted vector; **H** := channel matrix; **w** := AWGN **Sphere detector** finds:

$$\hat{\mathbf{x}} = \underset{\mathbf{x}\in\mathbf{Z}^n}{\operatorname{arg\,min}} \|\mathbf{H}\mathbf{x}-\mathbf{y}\|^2$$

 $\Box$  thus **H** is the lattice basis, **y** is the query point.

Block-based space-time decoding (Damen'2000 & many others) can be formulated in a similar way.

## CVP in Communications (2)

Sequence detection for ISI channels (Mow'91 & '94):

Sequence detector for ISI channels minimizes the metric:

 $\|\mathbf{H}\widetilde{\mathbf{u}}-(\mathbf{y}-\mathbf{G}\hat{\mathbf{u}})\|^2$ 

where  $\mathbf{y}$  := received seq.;  $\mathbf{G}$ ,  $\mathbf{H}$  := Toeplitz channel matrices;  $\tilde{\mathbf{u}}$  := integer-valued seq. to be detected;  $\hat{\mathbf{u}}$  := previously detected seq.

Let  $\mathbf{q} = \mathbf{y} - \mathbf{G}\hat{\mathbf{u}}$  be the query point, **H** be the lattice basis



 CDMA multiuser detection (Brunel et al.'98 &'2003) and MIMO sequence detection (Vikalo-Hassibi'2002) can be formulated similarly. presented by Wai Ho Mow

EEE, HKUST

(w.mow@ieee.org)

## Solving CVP approximately

- Very efficient algorithms exist for some special types of lattices, e.g. cubic lattices
- But in general, solving CVP is hard
- Solving CVP approximately is less difficult:
  - Nulling & rounding/quantization (*zero-forcing*)
  - Babai's nearest plane algorithm (DFE: successive nulling & cancellation)
  - Nearest plane algorithm with optimal ordering (VBLAST)
- These sub-optimal detectors have much lower complexity than the optimal MLD.

### Solving CVP exactly

- How to solve CVP exactly for optimal performance?
  Solving CVP in 2 steps:
  - 1. (*lattice reduction*) for a given lattice, find a "short" and fairly "orthogonal" basis
  - 2. (*sphere decoding*) enumerate all lattice points inside a sphere centered at the query point
- Lattice reduction can also enhance the performance of suboptimal schemes mentioned before

#### Lattice Reduction

The definition of basis reduction is not unique:

 for 2-D lattices: Gauss reduction
 Minkowski reduction: the shortest possible basis
 Lenstra, Lenstra & Lovász (LLL or L<sup>3</sup>) reduction

 LLL reduction is very important and useful in practical applications (such as cryptanalysis) as its complexity is only polynomial time

## Lattice Reduction Algorithm (1)

The first step of the reduction algorithm is called size-reduction:
[1]

$$\mathbf{b}_{i} \leftarrow \mathbf{b}_{i} - \sum_{j=i+1}^{m} \mu_{ij} \mathbf{b}_{j} \qquad \mu_{ij} = \left[ \frac{\left\langle \mathbf{b}_{i}, \mathbf{b}_{j}^{*} \right\rangle}{\left\| \mathbf{b}_{j}^{*} \right\|^{2}} \right]$$

- This operation shortens the lengths of the basis vectors (hence its name)
- After size reduction,  $|\mu_{i+1,i}| \le 0.5$

## Lattice Reduction Algorithm (2)

- Can we do better?
- We can re-order (how?) the basis vectors to obtain a new set of {b<sub>i</sub><sup>\*</sup>}
- After this re-ordering, | μ<sub>i+1,i</sub> | > 0.5, so perform the size-reduction again
- Repeat until no further improvement is achievable

#### Lattice Reduction Algorithm (3)

#### • For LLL, swap $\mathbf{b}_i$ and $\mathbf{b}_{i+1}$ if:

$$\delta \left\| \mathbf{b}_{i}^{*} \right\|^{2} > \left\| \mathbf{b}_{i+1}^{*} \right\|^{2} + \mu_{i+1,i}^{2} \left\| \mathbf{b}_{i}^{*} \right\|^{2} \qquad \frac{2}{3} < \delta < 1$$

- $\delta$  is a parameter:
  - $\Box$  choose small  $\delta$  for faster convergence
  - $\Box$  choose large  $\delta$  for better reduced basis
  - □ the basis might not be reduced and the algorithmic complexity is not polynomial time, if  $\delta$  is chosen outside the specified range.

#### Properties of LLL-reduced basis

- Denote  $\lambda(L)$  as the length of the shortest vector in L, then (when  $\delta = \frac{3}{4}$ ):
  - 1.  $||\mathbf{b}_1|| \le 2^{(n-1)/2} \lambda(L)$
  - 2.  $||\mathbf{b}_1|| \le 2^{(n-1)/4} \det(L)^{1/n}$
  - 3.  $||\mathbf{b}_1|| \dots ||\mathbf{b}_n|| \le 2^{n(n-1)/4} \det(L)$
- (1) & (2) ensure that the reduced-basis contains short vectors.
- (3) ensures a "near-orthogonal" basis.
- From experience these bounds are quite loose, i.e., the algorithm does better in practice.

# Sphere Decoding (1)

- Originally developed by Pohst in 1981
- To enumerate the lattice points inside a sphere centered at the query point
- A lattice point is identified coordinate by coordinate.



# Sphere Decoding (2)

- Identify the range of m-D "planes" (i.e. lines here) bounded by the sphere
- 2. Choose one of the plane within this range, then project the sphere onto this plane
- Identify the range of (m-1)-D "planes" (i.e. points) bounded by this new lower-dimensional "sphere", and choose one within the range
- 4. Perform the above recursively until eventually a lattice point is found, its distance to the query point can be calculated conveniently.
- 5. The radius of the sphere and the search ranges could be shrunk
- 6. Repeat until no more point can be found inside the sphere.



## Sphere Decoding (3)

- Sphere decoding can also be viewed as a *tree search*
- The radius defines a lower upper bounds for each dimension (level)
- Once a leaf node (i.e. lattice point) is reached, the bounds are updated
- The tree becomes smaller and smaller until no leaf can be found with the most current bounds.



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## Sphere Decoding Complexity

The SD complexity is sensitive to :

 the initial radius of the sphere
 the enumeration order

 Reduced basis may also lower the SD complexity

#### Packing Radius Test

- A simplification of the sphere decoder is to make use of the *packing radius* in a *sufficiency test*.
- Already used in Mow'91 & '94.
- A lattice point found inside the packing sphere of the query point must be the closest one, so the enumeration can be terminated immediately
- The packing radius which is a property of the lattice, can be found in the preprocessing stage and it needs to be updated only when the channel matrix requires so.

#### **BER Performance**

- Simulation of a 2transmitter 3-receiver MIMO system using 4-PAM
- Equivalent to solving CVP of 2D lattice in 6D space
- LLL reduction enhances the performance of various suboptimal schemes
- MLD performance was achieved by sphere decoding



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#### **BER Performance**

- Simulation of a 4transmitter-4-receiver
   MIMO system using 64-QAM
- Again LLL reduction enhances the performance of various suboptimal schemes
- Complex lattice based detectors (CLLL-ZF etc.) can provide the same performance



(w.mow@ieee.org)

### **MLD** Complexity Comparison

- Simulation of a 4-tx 4-rx MIMO flat fading channel with 64-QAM
- Time complexity is measured in *average CPU time per symbol*
- SD-Pohst: sphere decoder with the original Pohst ordering
- SE: sphere decoder with Schnorr-Euchner ordering



### **MLD** Complexity Comparison

- Simulation of a 4transmitter-4-receiver
   MIMO system using 64-QAM
- Time complexity is measured by the complexity exponent: log<sub>m</sub>(average #flops)
- Dotted line: sphere decoder without packing radius test
- Solid lines: sphere decoder with packing radius test



## **Concluding Remarks**

- Many communications detection problems can be reformulated as a CVP, so that the SD is applicable.
- The first of such communications detection problems solved is probably the MLSD problem for ISI channels.
- Lattice basis reduction is a powerful technique for improving the performance of various known algorithms (e.g. ZF, DFE, VBLAST) at the expense of higher preprocessing complexity.
- The packing radius test is an effective technique for reducing the average complexity (or power consumption) of MLD at the expense of higher preprocessing complexity.

## Concluding Remarks (2)

- The rich lattice/communications theory guarantees that many lattice related ideas are still waiting for us to explored!
- We have plenty of rooms for collaborations!!!
- To probe further:
  - W.H. Mow, "Universal Lattice Decoding: Principle and Recent Advances", Wireless Communications and Mobile Computing, Special Issue on Coding and Its Applications in Wireless CDMA Systems, Vol.3, Issue 5, August 2003, pp. 553-569.
  - □ http://www.ee.ust.hk/~eewhmow
- Finally, you might not know the impact of your present work until 10 years later!!!

#### Thank You