

## LLL lattice reduction

$L \subseteq \mathbb{R}^n$  full rank lattice.

$(b_1, \dots, b_n)$  ordered basis for  $L$ .

Associated to  $b_1, \dots, b_n$  is

unit vectors  $u_1, \dots, u_n$ , orthonormal  
s.t.  $\begin{bmatrix} b_1 & b_2 & \dots & b_n \end{bmatrix}$  when viewed

in this coordinate system, is  
upper triangular.

$b_i^* = b_i - \text{projection}(b_i) \text{ onto span}(b_1, \dots, b_{i-1})$

$$u_i = \frac{b_i^*}{\|b_i^*\|}$$

$$A(b_1, \dots, b_n) = \begin{bmatrix} \|b_1^*\| & & & \\ & \|b_2^*\| & & \\ & & \dots & \\ & & & \|b_n^*\| \end{bmatrix}$$


...

$$\begin{bmatrix} 0 & \dots & \dots \\ \vdots & \ddots & \vdots \\ \vdots & \vdots & \dots \end{bmatrix}$$

column  $i$  is  $b_i$  written  
in coordinates  $(u_1, \dots, u_m)$

Weakly reduced:  $|A_{ij}| \leq \frac{\|b_i^*\|}{2} \cdot \forall i < j$

LLL-reduced: Weakly reduced and

for each  $i$ , let  $\lambda, \lambda'$  be the  
projections of  $b_i, b_{i+1}$  on span  
of  $u_i, u_{i+1}$ .

want  $\|\lambda'\| \geq \underline{0.9} \|\lambda\|$   $\textcircled{*}$

[ Consider lattice  $L^*$  generated by  $\lambda, \lambda'$

in  $\mathbb{R}^c \approx \text{span}(u_i, u_{i+1})$ ,  
 then  $(\lambda, \lambda')$  being weakly reduced  
 and  $\|\lambda'\| \geq \|\lambda\| \iff \lambda$  is shortest  
 vector in  $L^*$

Thm Can find LLL reduced bases  
 in  $\text{poly}(n, t)$  time. (where all  
 entries of the given basis of the lattice  
 are  $st$ -bit integers).

Thm If  $(b_1, \dots, b_n)$  is LLL-reduced  
 then  $\|b_i\| \leq 2^{O(n)}$ . shortest  
 vector in  $L$ .

Thm If  $(b_1, \dots, b_n)$  is LLL-reduced  
 then  $\text{OD}(b_1, \dots, b_n) \leq 2^{O(n^2)}$ .

Thm Given a basis  $\{b_1, \dots, b_n\}$  with  $OD(b_1, \dots, b_n) \leq C$ ,  
We can find a shortest vector  
in time  $\text{poly}(n, t, C^n)$ .

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### LLL algorithm

Start with initial basis  $(b_1, \dots, b_n)$

While not LLL reduced, do:

Make  $(b_1, \dots, b_n)$  weakly reduced

If for some  $i$ ,  $b_i, b_{i+1}$  violate

$(*)$

swap  $b_i, b_{i+1}$ .

Done

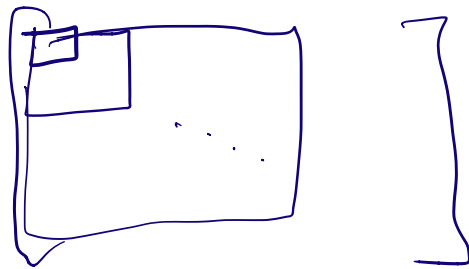
### Analysis

If algorithm stops, the result is

LLL-reduced.

Define potential function.  $\Phi(b_1, \dots, b_n)$   
as follows:

Take  $A(b_1, \dots, b_n)$



$$\Phi(b_1, \dots, b_n) = \prod_{j=1}^n \det \left( A_{[j] \times [j]}(b_1, \dots, b_n) \right)$$

$$= \prod_{j=1}^n \prod_{l=1}^j \|b_l^*\|$$

$$= \prod_{i=1}^n \|b_i^*\|^{n-i+1}$$

Claim Weak reduction does not change  $\Phi$ .

(Weak reduction does not change  $u_1, \dots, u_m$  and the diagonal of  $A$ ).



Claim If  $b_i, b_{i+1}$  violate  $(*)$ , then swapping  $b_i, b_{i+1}$  reduces  $\Phi$  by a constant factor.

Proof If we swap  $b_i, b_{i+1}$ , then  $b_1^*, b_2^*, \dots, b_{i-1}^*, b_{i+2}^*, \dots, b_n^*$  are unchanged.

Only  $b_i^*, b_{i+1}^*$  change.

Condition  $(*)$  violated



$$\begin{bmatrix} \|b_i^*\| & A_{i,i+1} \\ & \|b_{i+1}^*\| \end{bmatrix}$$

$$\|b_{i+1}^*\|^2 + (A_{i,i+1})^2 < 0.9 \cdot \|b_i^*\|^2$$

After swap:

$\left( \begin{array}{c} A_{i,i+1} \\ \|b_{i+1}^*\| \end{array} \right)$  is first vector,

$\left( \begin{array}{c} \|b_i^*\| \\ 0 \end{array} \right)$  is the second vector.

The new  $\|b_i^*\|$  equals the length of  $\left( \begin{array}{c} A_{i,i+1} \\ \|b_{i+1}^*\| \end{array} \right) < 0.9 \|b_i^*\|$   
old.

$$\text{The } \underline{\text{new}} \quad \|b_{i+1}^*\| = \frac{\text{Old } \|b_i^*\| \cdot \text{Old } \|b_{i+1}^*\|}{\text{New } \|b_i^*\|}.$$

$$\text{New } \|b_i^*\| < 0.9 \cdot \text{old } \|b_i^*\|$$

$$\text{New } \|b_{i+1}^*\| \cdot \text{New } \|b_i^*\| = \text{old } \|b_{i+1}^*\| \cdot \text{old } \|b_i^*\|$$

$\Rightarrow \Phi(b_1, \dots, b_n)$  reduced by a factor 0.9

$\Phi$  starts off  $< 2^{\text{poly}(n,t)}$

$\Phi$  is always integer  
 So  $\Phi \geq 1$ .

$$\begin{bmatrix} 1 \\ \vdots \\ b_i \\ \vdots \\ b_n \\ \vdots \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & & & & \\ & \ddots & & & \\ & & u_i & & \\ & & & \ddots & \\ & & & & 1 \end{bmatrix} \begin{bmatrix} a \\ \vdots \\ 0 \quad A \quad \vdots \\ \vdots \end{bmatrix}$$

$\parallel B$                        $\cup$                        $A$

$$B^T B = A^T A \cdot \begin{bmatrix} \square & \\ & \square \end{bmatrix} \cdot \begin{bmatrix} \square & \\ & \square \end{bmatrix}$$



Upper left  $j \times j$  submatrix of  
 $A^T A$

$$= \left( \text{upper left } j \times j \text{ submatrix of } A^T \right)$$

$$\times \left( \text{upper left } j \times j \text{ matrix of } A \right)$$

(since  $A$  is  
upper triangular)

$$= \det \left( A_{[j] \times [j]} \right)^2$$

$$\Phi(b_1, \dots, b_n)^2 = \prod_{j=1}^n \det \left( (B^T B)_{[j] \times [j]} \right)$$

$\in \mathbb{Z}$ .

[Did not prove: all intermediate  
integers are  $\leq \text{poly}(n, t)$  bits long].

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Proof that LL-reduced bases are nice.

Condition  $(*)$

$$\Rightarrow \|b_{i+1}^*\|^2 + |A_{i,i+1}|^2 \geq 0.9 \cdot \|b_i^*\|^2$$

Weak-reducedness

$$\Rightarrow |A_{i,i+1}| \leq \frac{\|b_i^*\|}{2}$$

$$\text{So } 0.9 \|b_i^*\|^2 \leq \|b_{i+1}^*\|^2 + \frac{\|b_i^*\|^2}{4}$$

$$\Rightarrow \|b_i^*\|^2 \leq \frac{\|b_{i+1}^*\|^2}{0.9 - 0.25}$$

$$\leq 2 \|b_{i+1}^*\|^2$$

So  $\|b_i^*\|$ 's are not decreasing too rapidly as  $i$  increases

$$\begin{aligned}
\|\text{Shortest vector}\| &\geq \min_i \|b_i^*\| \\
&\geq 2^{-n/2} \|b_1^*\| \\
&= 2^{-n/2} \|b_1\|.
\end{aligned}$$

So  $b_1$  is an approximate shortest vector.

~~Want~~ Analysis of OD.

$$\text{Want to say: } \frac{\prod \|b_i\|}{\prod \|b_i^*\|} \leq 2^{O(n^2)}$$

$$\|b_i\|^2 = \left( \sum_{l < i} |A_{li}|^2 \right) + \|b_i^*\|^2.$$

$$\left[ \begin{array}{c} i \\ \vdots \\ A_{li} \\ \vdots \\ \|b_i^*\| \\ \vdots \\ 0 \end{array} \right]$$

$$\leq \sum_{l < i} \frac{\|b_l^*\|^2}{4} + \|b_i^*\|^2$$

$$\leq \sum_{l < i} \left( \frac{\|b_l^*\|^2 \cdot 2^{i-l}}{4} \right) + \|b_i^*\|^2$$

$$\leq 2^n \cdot \|b_i^*\|^2$$

So done.

Finding exact shortest vectors.

Given  $\begin{bmatrix} b_1 & \dots & b_n \\ | & & | \\ \hline & & \end{bmatrix}$ , want to

minimize  $\|\sum x_i b_i\|$  as  $(x_1, \dots, x_n)$

vary in  $\mathbb{Z}^n$ .

Let  $v$  be the shortest vector in  $L$ .

Solution  $x$  to

$$\begin{bmatrix} b_1 \\ \vdots \\ b_i \\ \vdots \\ b_n \end{bmatrix} (x) = (v)$$

is given by Cramer's rule,

$$x_i = \frac{\det \begin{pmatrix} b_1, \dots, b_{i-1}, v, b_{i+1}, \dots, b_n \end{pmatrix}}{\det \begin{pmatrix} b_1, \dots, b_n \end{pmatrix}}$$

$$\leq \frac{\prod_{\substack{j=1 \\ j \neq i}}^n \|b_j\| \cdot \|v\|}{\det \begin{pmatrix} b_1, \dots, b_n \end{pmatrix}}$$

$$\leq \frac{\prod_j \|b_j\|}{\det(b_1, \dots, b_n)} \leq OD(b_1, \dots, b_n)$$

Algorithm try all  
 $(x_1, \dots, x_n) \in [-OD, OD]^n$   
 and output  $\sum x_i b_i$   
 with smallest norm.

So in time  $\text{poly}(2^{\text{poly}(n)}, t)$ ,  
 we can find exact  
 shortest vectors in  
 lattices.

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