

# Motivation and applications of LLL

HATS lightning talk

April 25, 2020

# Lattices

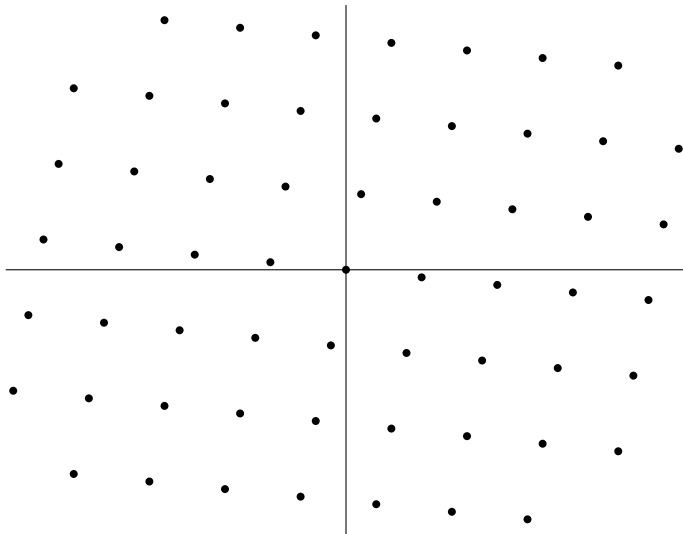
Let  $\mathbf{b}_i$  be some vectors, the lattice generated by  $\mathbf{b}_i$  is  $\sum_i \mathbb{Z}\mathbf{b}_i$ , containing elements like

$$1\mathbf{b}_1 + 4\mathbf{b}_2 - 3\mathbf{b}_3$$

$$-\mathbf{b}_2 + 2\mathbf{b}_3$$

$$\mathbf{b}_1$$

# Lattice in $\mathbb{R}^2$



# Euclidean algorithm

The Euclidean algorithm returns the gcd of  $a, b$

```
while  $b \neq 0$  do  
  if  $|a| > |b|$  then  
     $a, b \leftarrow b, a$   
  end if  
   $d \leftarrow \frac{b}{a}$   
   $b \leftarrow b - \lfloor d \rfloor a$   
end while  
return  $a$ 
```

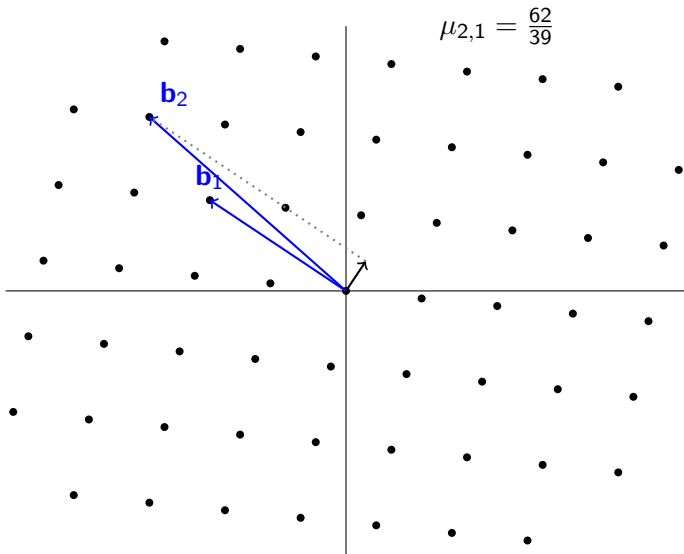
$a, b$  is just a lattice in  $\mathbb{R}^1$  and  $\gcd(a, b)$  is it's reduced lattice

# Gaussian Lattice Reduction

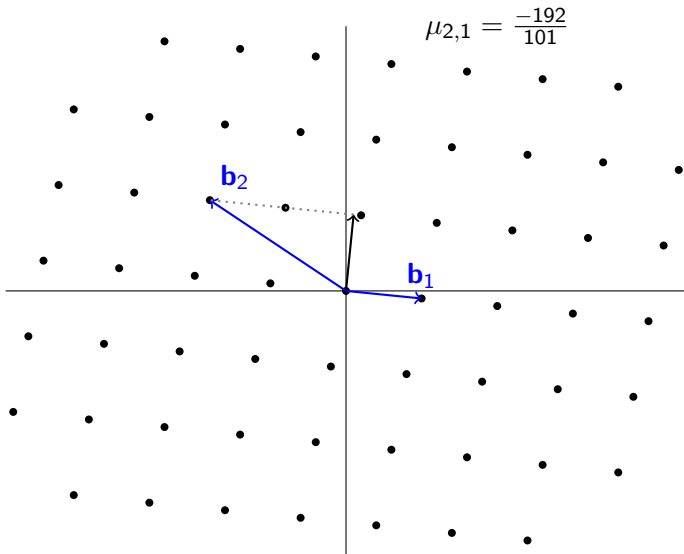
Let  $\mathbf{b}_1, \mathbf{b}_2$  be a basis.

```
while  $\lfloor \mu_{2,1} \rfloor \neq 0$  do  
  if  $\|\mathbf{b}_1\| > \|\mathbf{b}_2\|$  then  
     $\mathbf{b}_1, \mathbf{b}_2 \leftarrow \mathbf{b}_2, \mathbf{b}_1$   
  end if  
   $\mu_{2,1} \leftarrow \frac{(\mathbf{b}_2, \mathbf{b}_1)}{\|\mathbf{b}_1\|^2}$   
   $\mathbf{b}_2 \leftarrow \mathbf{b}_2 - \lfloor \mu_{2,1} \rfloor \mathbf{b}_1$   
end while  
return  $\mathbf{b}_1, \mathbf{b}_2$ 
```

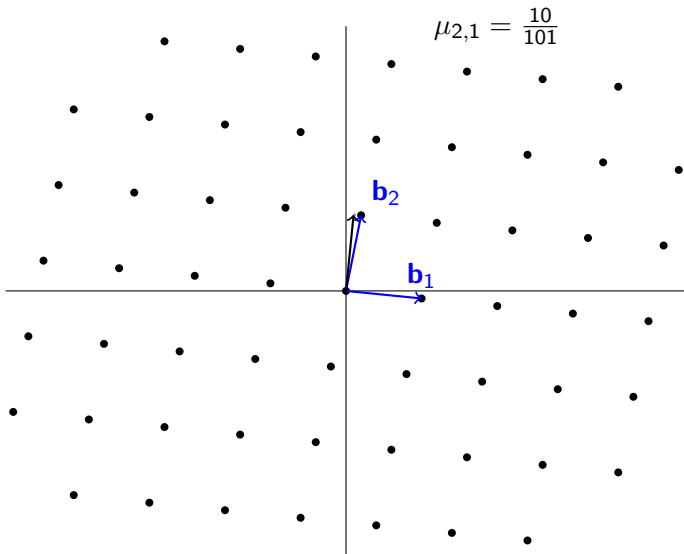
# Example



# Example



# Example





# Gram-Schmidt

For some vectors  $\mathbf{b}_i \in \mathbb{R}^n$ , define the orthogonal vectors  $\mathbf{b}_i^*$  as

$$\mathbf{b}_i^* = \mathbf{b}_i - \sum_{j=1}^{i-1} \frac{(\mathbf{b}_i, \mathbf{b}_j^*)}{\|\mathbf{b}_j^*\|^2} \mathbf{b}_j^* = \mathbf{b}_i - \sum_{j=1}^{i-1} \mu_{j,i} \mathbf{b}_j^*$$

with  $\mu_{i,j} = \frac{(\mathbf{b}_i, \mathbf{b}_j^*)}{\|\mathbf{b}_j^*\|^2}$

Then the space generated by  $\mathbf{b}_i$  and  $\mathbf{b}_i^*$  are the same. Typically we normalize the vectors but for lattice reduction purposes this is not done.

## LLL-reduced

For some basis  $\mathbf{b}_i$ , let  $\mathbf{b}_i^*$  be the Gram-Schmidt orthogonalized basis. Then the basis is LLL-reduced for  $\delta \in (\frac{1}{4}, 1)$  iff:

1. Size reduced:  $j < i, \mu_{i,j} \leq \frac{1}{2}$
2. Lovász condition:  $(\delta - \mu_{i+1,i}^2) \|\mathbf{b}_i^*\|^2 \leq \|\mathbf{b}_{i+1}^*\|^2$

The Lovász condition in some sense is ensuring that the vectors are 'sufficiently' orthogonal. Note that the bound given by these are just theoretical, it is common for these to be overestimations.

## LLL algorithm

```
 $i \leftarrow 2$   
while  $i < n$  do  
  for  $j = i - 1, i - 2, \dots, 1$  do  
    if  $|\mu_{i,j}| > \frac{1}{2}$  then  
       $\mathbf{b}_i \leftarrow \mathbf{b}_i - \lfloor \mu_{i,j} \rfloor \mathbf{b}_j$   
    end if  
  end for  
  if  $\left( \delta - \mu_{i,i-1}^2 \right) \|\mathbf{b}_{i-1}^*\|^2 \leq \|\mathbf{b}_i^*\|^2$  then  
     $i \leftarrow i + 1$   
  else  
     $i \leftarrow \max(i - 1, 2)$   
     $\mathbf{b}_{i-1}, \mathbf{b}_i \leftarrow \mathbf{b}_i, \mathbf{b}_{i-1}$   
  end if  
end while
```

# Applications

1. Rational approximation
2. Algebraic number approximation
3. Small roots

## Rational approximation

To find a rational approximation of  $x$ , let  $B$  be a big number.

$$\begin{pmatrix} 1 & 0 & xB \\ 0 & 1 & -B \end{pmatrix}$$

Then the smallest vector of the LLL reduced matrix is of the form  $(a, b, k)$  where  $0 \approx \frac{k}{B} = a - bx$ , hence  $x \approx \frac{b}{a}$

## Algebraic number approximation

A algebraic number is a root to a polynomial with integer coefficients.

To find a algebraic approximation of  $x$ , let  $B$  be a big number and  $n$  be the degree of a polynomial.

$$\begin{pmatrix} 1 & 0 & \dots & 0 & B \\ 0 & 1 & \dots & 0 & xB \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & x^n B \end{pmatrix}$$

Then the smallest vector of the LLL reduced matrix is of the form  $(f_0, f_1, \dots, f_n, k)$  with  $k$  small and  $x$  is approximately the root of  $\sum f_i y^i$

# Howgrave Graham

Let  $f(x)$  be some univariate polynomial with  $n$  monomials. For some modulus  $M$  and bound  $B$ :

$f(x_0) \equiv 0 \pmod{M}$ ,  $x_0 < B$  and  $|f(x)| < M$  for all  $0 < x < B$  implies  $f(x_0) = 0$  over  $\mathbb{Z}$ .

$f(x_0) \equiv 0 \pmod{B}$  and  $\|f(Bx)\|_2 < \frac{B}{\sqrt{n}}$  implies  $f(x_0) = 0$  over  $\mathbb{Z}$ .

## Coppersmith algorithm

If  $x_0 < B$  and  $f(x_0) \pmod{N} = 0$  for some polynomial  $f(x) = \sum_{i=0}^n f_i x^i$  with  $f_n = 1$ , then construct the lattice

$$\begin{pmatrix} N & 0 & 0 & \dots & 0 & 0 \\ 0 & NB & 0 & \dots & 0 & 0 \\ 0 & 0 & NB^2 & \dots & 0 & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & 0 & \dots & NB^{n-1} & 0 \\ f_0 & f_1 B & f_2 B^2 & \dots & f_{n-1} B^{n-1} & f_n B^n \end{pmatrix}$$

If  $\mathbf{v}$  is the shortest vector after LLL, then  $g(x) = \sum_{i=0}^n v_i x^i$  possibly a root  $\frac{x_0}{B}$  over the reals.



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