

# **Short Vectors of Planar Lattices via Continued Fractions**

**Friedrich Eisenbrand**

**Max-Planck-Institut für Informatik**

# Outline

## Lattices

- Definition of planar integral lattices
- Shortest Vectors
- Applications
- The Gaussian algorithm

# Outline

## Lattices

- Definition of planar integral lattices
- Shortest Vectors
- Applications
- The Gaussian algorithm

## Short vectors and best approximations

- Hermite normal form
- Shortest vectors and best approximations
- Best approximations and convergents
- A naive shortest vector algorithm based on euclid

# Outline

## Lattices

- Definition of planar integral lattices
- Shortest Vectors
- Applications
- The Gaussian algorithm

## Short vectors and best approximations

- Hermite normal form
- Shortest vectors and best approximations
- Best approximations and convergents
- A naive shortest vector algorithm based on euclid

## Fast basis reduction

- Identifying the “shortest convergent”
- Computing shortest convergent w.r.t.  $\ell_\infty$ -norm with Schönages speedup
- Reduced bases
- Fast basis reduction via (Schönhage 1971) and (Gauß 1801)

# Lattices

**Lattice** generated by  $A$ :  $\Lambda(A) = \{Ax \mid x \in \mathbf{Z}^2\}$ ,

where  $A \in \mathbf{Z}^{2 \times 2}$  nonsingular

# Lattices

Lattice generated by  $A$ :  $\Lambda(A) = \{Ax \mid x \in \mathbf{Z}^2\}$ ,

where  $A \in \mathbf{Z}^{2 \times 2}$  nonsingular

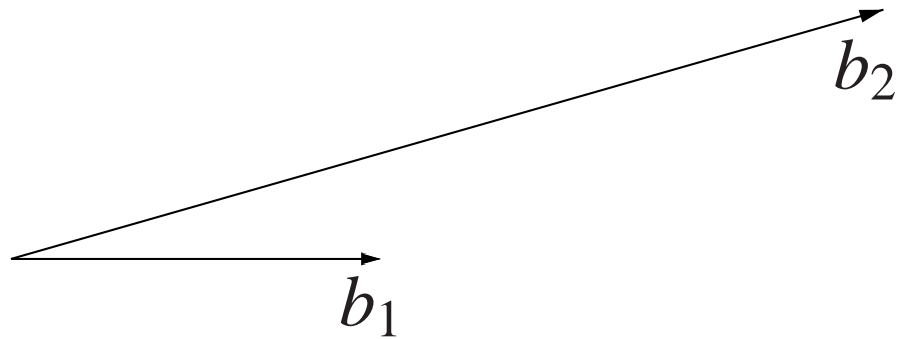
$A$  basis of  $\Lambda$

# Lattices

**Lattice** generated by  $A$ :  $\Lambda(A) = \{Ax \mid x \in \mathbf{Z}^2\}$ ,

where  $A \in \mathbf{Z}^{2 \times 2}$  nonsingular

$A$  **basis** of  $\Lambda$

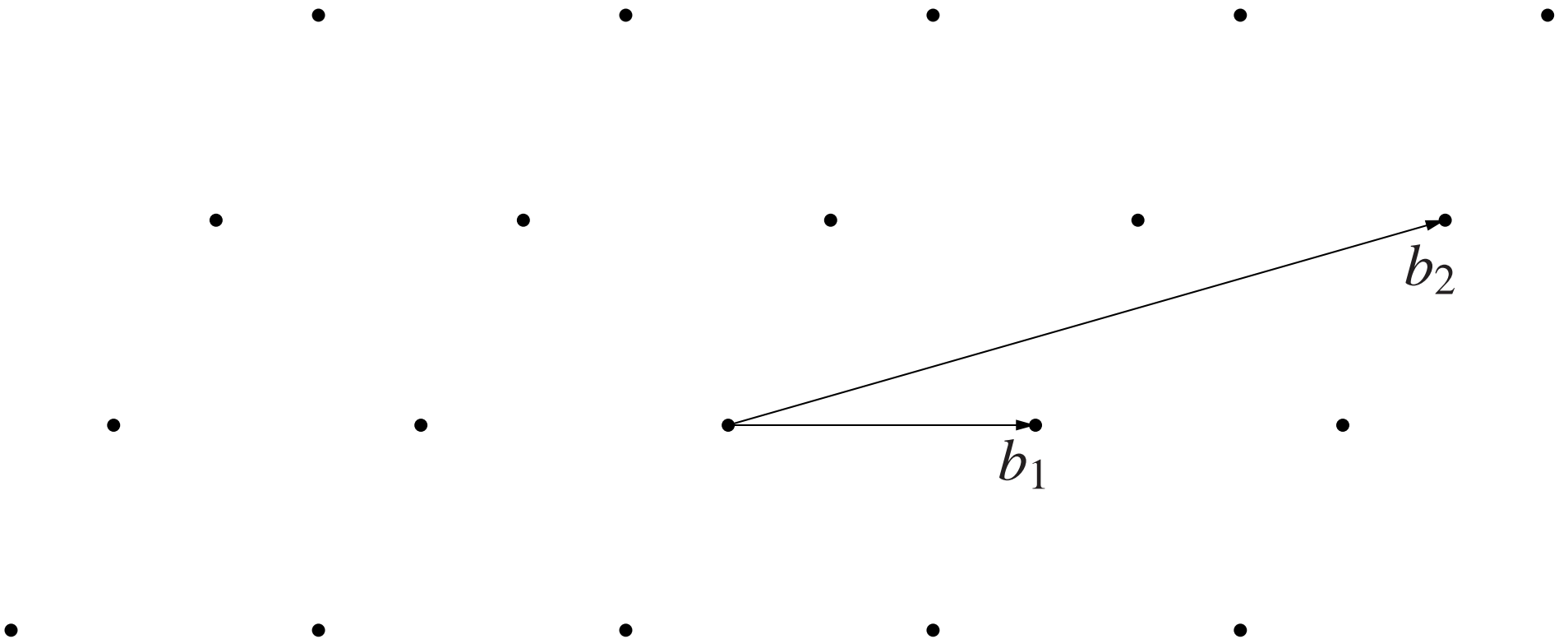


# Lattices

Lattice generated by  $A$ :  $\Lambda(A) = \{Ax \mid x \in \mathbf{Z}^2\}$ ,

where  $A \in \mathbf{Z}^{2 \times 2}$  nonsingular

A basis of  $\Lambda$





## Shortest vectors

Shortest vector of  $\Lambda$ :  $v \in \Lambda$  with  $v \neq 0$  and  $\|v\| \leq \|u\|$  for all  $u \in \Lambda$   
 $u \neq 0$

# Shortest vectors

**Shortest vector of  $\Lambda$ :**  $v \in \Lambda$  with  $v \neq 0$  and  $\|v\| \leq \|u\|$  for all  $u \in \Lambda$   
 $u \neq 0$

## Applications

- Integer programming in fixed dimension Lenstra (1983)
- Factorization of rational polynomials Lenstra, Lenstra & Lovász (1982)
- Strongly polynomial time algorithms in combinatorial optimization Frank & Tardos (1987)

# Shortest vectors

**Shortest vector of  $\Lambda$ :**  $v \in \Lambda$  with  $v \neq 0$  and  $\|v\| \leq \|u\|$  for all  $u \in \Lambda$   
 $u \neq 0$

## Applications

- Integer programming in fixed dimension Lenstra (1983)
- Factorization of rational polynomials Lenstra, Lenstra & Lovász (1982)
- Strongly polynomial time algorithms in combinatorial optimization Frank & Tardos (1987)

# Shortest vectors

**Shortest vector of  $\Lambda$ :**  $v \in \Lambda$  with  $v \neq 0$  and  $\|v\| \leq \|u\|$  for all  $u \in \Lambda$   
 $u \neq 0$

## Applications

- Integer programming in fixed dimension Lenstra (1983)
- Factorization of rational polynomials Lenstra, Lenstra & Lovász (1982)
- Strongly polynomial time algorithms in combinatorial optimization Frank & Tardos (1987)

# Unimodular transformations

Unimodular transformation of basis  $A$ :

$$A \cdot U,$$

where  $U \in \mathbf{Z}^{2 \times 2}$  with  $\det(U) = \pm 1$ , .i.e.,  $U$  unimodular

# Unimodular transformations

Unimodular transformation of basis  $A$ :

$$A \cdot U,$$

where  $U \in \mathbf{Z}^{2 \times 2}$  with  $\det(U) = \pm 1$ , .i.e.,  $U$  unimodular

Matrix  $A$  basis of  $\Lambda$  **if and only if**  $A \cdot U$  basis of  $\Lambda$ ,  $U$  unimodular

# Unimodular transformations

Unimodular transformation of basis  $A$ :

$$A \cdot U,$$

where  $U \in \mathbf{Z}^{2 \times 2}$  with  $\det(U) = \pm 1$ , .i.e.,  $U$  unimodular

Matrix  $A$  basis of  $\Lambda$  **if and only if**  $A \cdot U$  basis of  $\Lambda$ ,  $U$  unimodular

## Examples of unimodular transformations

- swapping of columns

# Unimodular transformations

Unimodular transformation of basis  $A$ :

$$A \cdot U,$$

where  $U \in \mathbf{Z}^{2 \times 2}$  with  $\det(U) = \pm 1$ , .i.e.,  $U$  unimodular

Matrix  $A$  basis of  $\Lambda$  **if and only if**  $A \cdot U$  basis of  $\Lambda$ ,  $U$  unimodular

## Examples of unimodular transformations

- swapping of columns
- adding integral multiples of one column to another



# The Gaussian reduction algorithm

GAUSS( $b_1, b_2$ )

**repeat**

arrange that  $b_1$  is the shorter vector of  $b_1$  and  $b_2$

find  $k \in \mathbf{Z}$  such that  $b_2 - kb_1$  is of minimal euclidean length

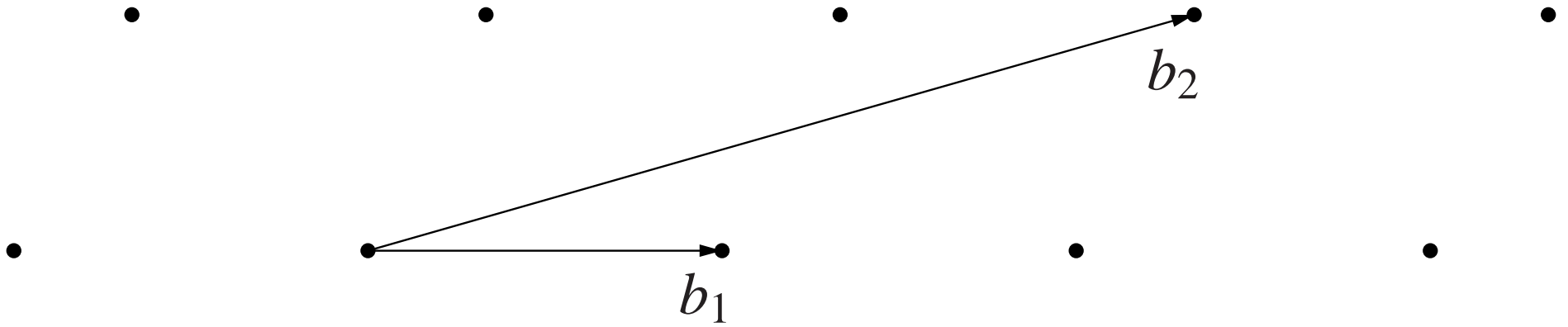
$b_2 \leftarrow (b_2 - kb_1)$  (normalization step)

**until**  $k = 0$

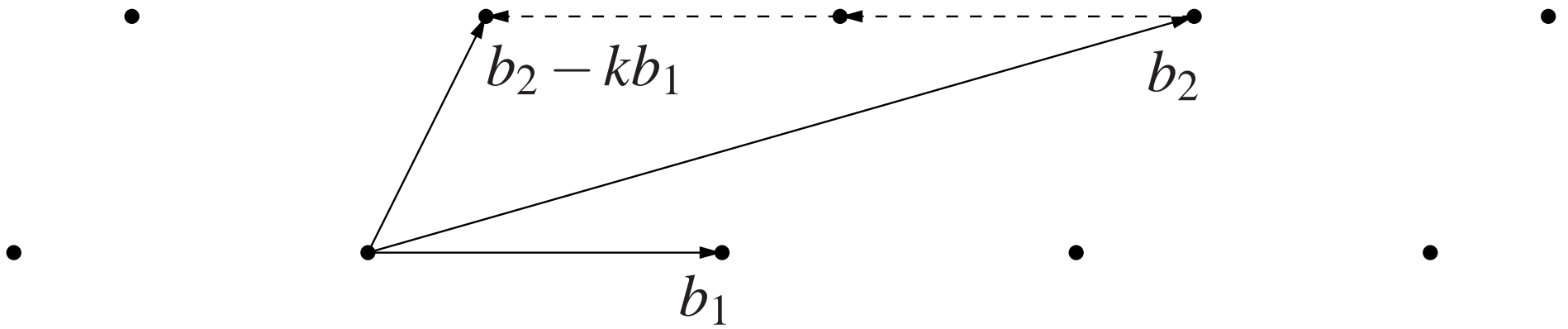
**return** ( $b_1, b_2$ )

$k$  in **repeat**-loop is nearest integer to  $(b_1^T b_2) / (b_1^T b_1)$

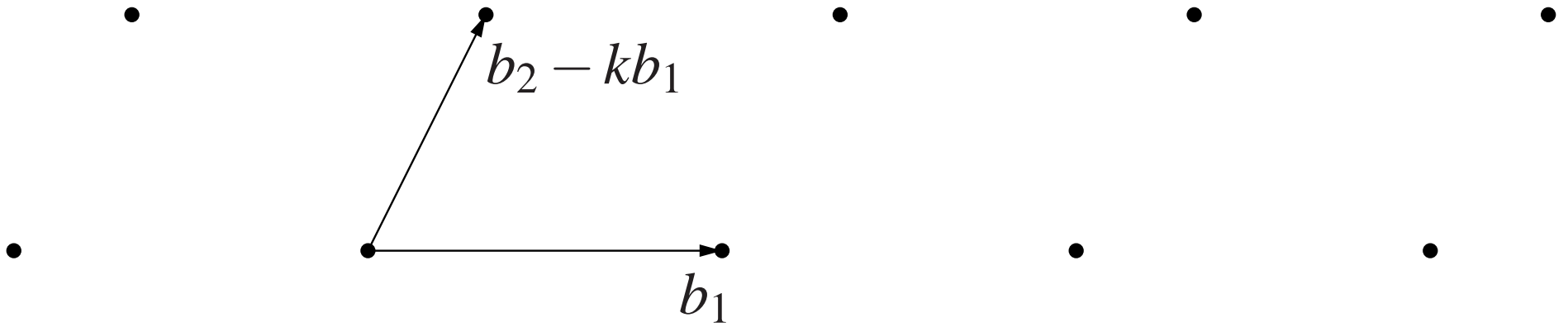
# A normalization step



# A normalization step



# A normalization step



# Complexity of GAUSS

- Lagarias (1980): Complexity of GAUSS  $O(n^3)$

# Complexity of GAUSS

- Lagarias (1980): Complexity of GAUSS  $O(n^3)$
- Schönhage (1991) and Yap (1992) new reduction algorithm:  
Complexity  $O(M(n) \log n)$  where  $M(n)$ : complexity of  $n$ -bit integer multiplication;

# Complexity of GAUSS

- Lagarias (1980): Complexity of GAUSS  $O(n^3)$
- Schönhage (1991) and Yap (1992) new reduction algorithm:  
Complexity  $O(M(n) \log n)$  where  $M(n)$ : complexity of  $n$ -bit integer multiplication;  
algorithms fairly involved

# Complexity of GAUSS

- Lagarias (1980): Complexity of GAUSS  $O(n^3)$
- Schönhage (1991) and Yap (1992) new reduction algorithm:  
Complexity  $O(M(n) \log n)$  where  $M(n)$ : complexity of  $n$ -bit integer multiplication;  
algorithms fairly involved
- **This talk:** Fast basis reduction via Schönhage's (1971) classical gcd-speedup and GAUSS



# The Hermite Normal Form

Given lattice basis  $A = \begin{pmatrix} a_1 & a_2 \\ a_3 & a_4 \end{pmatrix} \in \mathbf{Z}^{2 \times 2}$ , compute integers  $x, y$  with  $\gcd(a_3, a_4) = d = xa_3 + ya_4$

$\begin{pmatrix} a_4/d & x \\ -a_3/d & y \end{pmatrix}$  unimodular

# The Hermite Normal Form

Given lattice basis  $A = \begin{pmatrix} a_1 & a_2 \\ a_3 & a_4 \end{pmatrix} \in \mathbf{Z}^{2 \times 2}$ , compute integers  $x, y$  with  $\gcd(a_3, a_4) = d = xa_3 + ya_4$

$\begin{pmatrix} a_4/d & x \\ -a_3/d & y \end{pmatrix}$  unimodular

Then

$$\begin{pmatrix} a_1 & a_2 \\ a_3 & a_4 \end{pmatrix} \begin{pmatrix} a_4/d & x \\ -a_3/d & y \end{pmatrix} = \begin{pmatrix} a & b \\ 0 & c \end{pmatrix} \in \mathbf{Z}^{2 \times 2}$$

# The Hermite Normal Form

Given lattice basis  $A = \begin{pmatrix} a_1 & a_2 \\ a_3 & a_4 \end{pmatrix} \in \mathbf{Z}^{2 \times 2}$ , compute integers  $x, y$  with  $\gcd(a_3, a_4) = d = xa_3 + ya_4$

$\begin{pmatrix} a_4/d & x \\ -a_3/d & y \end{pmatrix}$  unimodular

Then

$$\begin{pmatrix} a_1 & a_2 \\ a_3 & a_4 \end{pmatrix} \begin{pmatrix} a_4/d & x \\ -a_3/d & y \end{pmatrix} = \begin{pmatrix} a & b \\ 0 & c \end{pmatrix} \in \mathbf{Z}^{2 \times 2}$$

We can assume that  $c > 0$  and  $a > b \geq 0$

Hermite normal form

## Complexity of Hermite normal form

Given integers  $a$  and  $b$  one can compute integers  $x$  and  $y$  with

$$\gcd(a, b) = xa + yb$$

in time  $O(M(n) \log n)$  (Schönhage 1971)

**Complexity of HNF:**  $O(M(n) \log n)$

# Complexity of Hermite normal form

Given integers  $a$  and  $b$  one can compute integers  $x$  and  $y$  with

$$\gcd(a, b) = xa + yb$$

in time  $O(M(n) \log n)$  (Schönhage 1971)

**Complexity of HNF:**  $O(M(n) \log n)$

**Assuming:** Lattice is given by its Hermite normal form

# Best approximations

For  $\alpha \in \mathcal{Q}$ , fraction  $x/y$  is **best approximation** if all other fractions  $x'/y'$  with  $y' \leq y$  satisfy

$$|y'\alpha - x'| > |y\alpha - x|$$

## Shortest vectors and best approximations

$$\Lambda = \left\{ \begin{pmatrix} a & b \\ 0 & c \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \mid x, y \in \mathbf{Z} \right\}$$

**Theorem.** *There exists a shortest vector  $\begin{pmatrix} -xa+yb \\ yc \end{pmatrix}$ ,  $x \in \mathbf{N}_0$ ,  $y \in \mathbf{N}_+$  of  $\Lambda$  such that at least one of the following conditions is satisfied.*

# Shortest vectors and best approximations

$$\Lambda = \left\{ \begin{pmatrix} a & b \\ 0 & c \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \mid x, y \in \mathbf{Z} \right\}$$

**Theorem.** *There exists a shortest vector  $\begin{pmatrix} -xa+yb \\ yc \end{pmatrix}$ ,  $x \in \mathbf{N}_0$ ,  $y \in \mathbf{N}_+$  of  $\Lambda$  such that at least one of the following conditions is satisfied.*

- *The fraction  $x/y$  is a best approximation to the number  $b/a$ .*



# Shortest vectors and best approximations

$$\Lambda = \left\{ \begin{pmatrix} a & b \\ 0 & c \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} \mid x, y \in \mathbf{Z} \right\}$$

**Theorem.** *There exists a shortest vector  $\begin{pmatrix} -xa+yb \\ yc \end{pmatrix}$ ,  $x \in \mathbf{N}_0$ ,  $y \in \mathbf{N}_+$  of  $\Lambda$  such that at least one of the following conditions is satisfied.*

- *The fraction  $x/y$  is a best approximation to the number  $b/a$ .*
- *If the fraction  $p/q$  is the reduced representation of  $b/a$ , then  $p$  is odd,  $q$  is even,  $x \in \{\lfloor p/2 \rfloor, \lceil p/2 \rceil\}$  and  $y = q/2$ .*

## Best approximations via euclidean algorithm

$x/y$  best approximation of  $b/a$ , then  $x/y$  is a **convergent** of  $b/a$

## Best approximations via euclidean algorithm

$x/y$  best approximation of  $b/a$ , then  $x/y$  is a **convergent** of  $b/a$

Convergents of  $\alpha \in \mathbf{Q}$  **inductively defined**

# Best approximations via euclidean algorithm

$x/y$  best approximation of  $b/a$ , then  $x/y$  is a **convergent** of  $b/a$

Convergents of  $\alpha \in \mathbf{Q}$  **inductively defined**

- $[\alpha]$

## Best approximations via euclidean algorithm

$x/y$  best approximation of  $b/a$ , then  $x/y$  is a **convergent** of  $b/a$

Convergents of  $\alpha \in \mathbf{Q}$  **inductively defined**

- $\lfloor \alpha \rfloor$
- $\lfloor \alpha \rfloor + 1/c$ , where  $c$  convergent of  $1/(\alpha - \lfloor \alpha \rfloor)$

# The Gregorian calendar

The earth turns around the sun in  $365 + 104\,629/432\,000$  days.

First convergent is 365

# The Gregorian calendar

The earth turns around the sun in  $365 + 104\,629/432\,000$  days.

First convergent is 365

Second convergent:  $365 + 1/4$ , since 4 is first convergent of  $432\,000/104\,629 = 4 + 13\,484/104\,629$

# The Gregorian calendar

The earth turns around the sun in  $365 + 104\,629/432\,000$  days.

First convergent is 365

Second convergent:  $365 + 1/4$ , since 4 is first convergent of  $432\,000/104\,629 = 4 + 13\,484/104\,629$

Leap year all 4 years



# The Gregorian calendar

Pope Gregor XIII (1582) used fifth convergent

$$365 + \frac{1}{4 + \frac{1}{7 + \frac{1}{1 + \frac{1}{3 + \frac{1}{6}}}}} = 365 + 194/801$$

# The Gregorian calendar

Pope Gregor XIII (1582) used fifth convergent

$$365 + \frac{1}{4 + \frac{1}{7 + \frac{1}{1 + \frac{1}{3 + \frac{1}{6}}}}} = 365 + 194/801$$

Every 800 years skip 6 leap years; every year divisible by 100 but not by 400

# The Gregorian calendar

Pope Gregor XIII (1582) used fifth convergent

$$365 + \frac{1}{4 + \frac{1}{7 + \frac{1}{1 + \frac{1}{3 + \frac{1}{6}}}}} = 365 + 194/801$$

Every 800 years skip 6 leap years; every year divisible by 100 but not by 400

The first year the in which calendar is 1 day ahead is 4915

# Computing convergents

## Euclidean algorithm

EXGCD( $a, b$ )

$$M \leftarrow \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$n \leftarrow 0$

**while** ( $b \neq 0$ ) **do**

$$q \leftarrow \lfloor a/b \rfloor$$

$$M \leftarrow M \begin{pmatrix} q & 1 \\ 1 & 0 \end{pmatrix}$$

$$(a, b) \leftarrow (b, a - qb)$$

$$n \leftarrow n + 1$$

**return** ( $d = a, x = (-1)^n M_{2,2}, y = (-1)^{n+1} M_{1,2}$ )

## Computing convergents

Let  $M^{(k)}$ ,  $k \geq 0$ , denote  $M$  after  $k$ -th iteration of while-loop

# Computing convergents

Let  $M^{(k)}$ ,  $k \geq 0$ , denote  $M$  after  $k$ -th iteration of while-loop

Well known fact:

$k$ -th convergent of  $a/b$  is  $M_{1,1}^{(k)} / M_{2,1}^{(k)}$

# Naive shortest vector algorithm

- Compute Hermite normal form  $\begin{pmatrix} a & b \\ 0 & c \end{pmatrix}$  of  $A$

## Naive shortest vector algorithm

- Compute Hermite normal form  $\begin{pmatrix} a & b \\ 0 & c \end{pmatrix}$  of  $A$
- Compute reduced fraction  $p/q$  of  $b/a$  and vectors  $(- \lfloor p/2 \rfloor a + \lfloor q/2 \rfloor b, \lfloor q/2 \rfloor c)^T$  and  $(a, 0)^T$  store shortest one in MIN



## Naive shortest vector algorithm

- Compute Hermite normal form  $\begin{pmatrix} a & b \\ 0 & c \end{pmatrix}$  of  $A$
- Compute reduced fraction  $p/q$  of  $b/a$  and vectors  $(-\lfloor p/2 \rfloor a + \lfloor q/2 \rfloor b, \lfloor q/2 \rfloor c)^T$  and  $(a, 0)^T$  store shortest one in MIN
- Compute convergents  $g_k/h_k$  of  $b/a$  with EXGCD( $b, a$ ) and compare the length of the induced vector  $(-g_k a + h_k b, h_k c)^T$  with MIN. If shorter, replace MIN

# Naive shortest vector algorithm

- Compute Hermite normal form  $\begin{pmatrix} a & b \\ 0 & c \end{pmatrix}$  of  $A$
- Compute reduced fraction  $p/q$  of  $b/a$  and vectors  $(-\lfloor p/2 \rfloor a + \lfloor q/2 \rfloor b, \lfloor q/2 \rfloor c)^T$  and  $(a, 0)^T$  store shortest one in MIN
- Compute convergents  $g_k/h_k$  of  $b/a$  with EXGCD( $b, a$ ) and compare the length of the induced vector  $(-g_k a + h_k b, h_k c)^T$  with MIN. If shorter, replace MIN

In the end MIN contains a shortest vector

## Naive shortest vector algorithm

- Compute Hermite normal form  $\begin{pmatrix} a & b \\ 0 & c \end{pmatrix}$  of  $A$
- Compute reduced fraction  $p/q$  of  $b/a$  and vectors  $(-\lfloor p/2 \rfloor a + \lfloor q/2 \rfloor b, \lfloor q/2 \rfloor c)^T$  and  $(a, 0)^T$  store shortest one in MIN
- Compute convergents  $g_k/h_k$  of  $b/a$  with EXGCD( $b, a$ ) and compare the length of the induced vector  $(-g_k a + h_k b, h_k c)^T$  with MIN. If shorter, replace MIN

In the end MIN contains a shortest vector

Linear search through all convergents of  $b/a$

## Finding shortest vector w.r.t $\ell_\infty$ -norm

Consider set of vectors

$$\left\{ \begin{pmatrix} -g_k a + h_k b \\ h_k c \end{pmatrix} \mid k = 0, \dots, t \right\}, \quad (1)$$

where  $\beta_k = g_k/h_k$ ,  $0 \leq k \leq t$  are the convergents of  $b/a$ .

**Proposition.** *Shortest vector in (1) w.r.t.  $\ell_\infty$  is last convergent of  $b/a$  outside the interval  $[(b-c)/a, (b+c)/a]$  or first convergent of  $b/a$  inside  $[(b-c)/a, (b+c)/a]$ .*

## Finding shortest vector w.r.t $\ell_\infty$ -norm

Common convergent of interval  $[\alpha_1, \alpha_2]$ : Convergent  $\beta_k$  of  $\alpha_1$  and  $\alpha_2$   
where  $k$  maximal

## Finding shortest vector w.r.t $\ell_\infty$ -norm

Common convergent of interval  $[\alpha_1, \alpha_2]$ : Convergent  $\beta_k$  of  $\alpha_1$  and  $\alpha_2$  where  $k$  maximal

Schönhage (1971): One can compute common convergent  $\beta_k$  and corresponding matrix  $M^{(k)}$  of two rationals  $\alpha_1, \alpha_2 \in \mathbf{Q}$  in time  $O(M(n) \log n)$

## Finding shortest vector w.r.t $\ell_\infty$ -norm

Common convergent of interval  $[\alpha_1, \alpha_2]$ : Convergent  $\beta_k$  of  $\alpha_1$  and  $\alpha_2$  where  $k$  maximal

Schönhage (1971): One can compute common convergent  $\beta_k$  and corresponding matrix  $M^{(k)}$  of two rationals  $\alpha_1, \alpha_2 \in \mathbf{Q}$  in time  $O(M(n) \log n)$

**Proposition.** *Let  $\beta_k = g_k/h_k$  common convergent of  $[(b-c)/a, (b+c)/a]$ . Then  $k$ -th,  $k+1$ -st or  $k+2$ -nd convergent of  $b/a$  is shortest vector in (1) w.r.t. the  $\ell_\infty$ -norm.*

# Fast shortest vector algorithm

- Compute HNF  $\begin{pmatrix} a & b \\ 0 & c \end{pmatrix}$  of  $A$



# Fast shortest vector algorithm

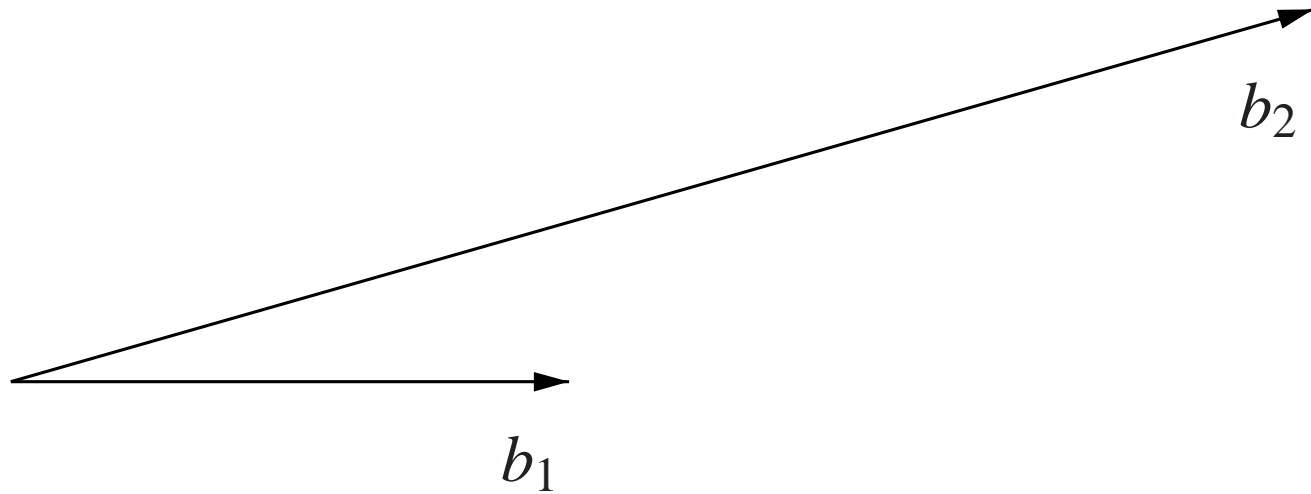
- Compute HNF  $\begin{pmatrix} a & b \\ 0 & c \end{pmatrix}$  of  $A$
- Compute reduced representation  $p/q$  of  $b/a$  and vectors  $(a, 0)^T$ ,  $(- \lfloor p/2 \rfloor a + \lfloor q/2 \rfloor b, \lfloor q/2 \rfloor c)^T$  ; store shortest nonzero one in a container MIN

# Fast shortest vector algorithm

- Compute HNF  $\begin{pmatrix} a & b \\ 0 & c \end{pmatrix}$  of  $A$
- Compute reduced representation  $p/q$  of  $b/a$  and vectors  $(a, 0)^T$ ,  $(-\lfloor p/2 \rfloor a + \lfloor q/2 \rfloor b, \lfloor q/2 \rfloor c)^T$ ; store shortest nonzero one in a container MIN
- Compute common convergent  $\beta_k$  of  $[(b - c)/a, (b + c)/a]$  and corresponding matrix  $M^{(k)}$ . Compute next two convergents of  $b/a$  with EXGCD; Replace MIN if one of convergents yields shorter vector

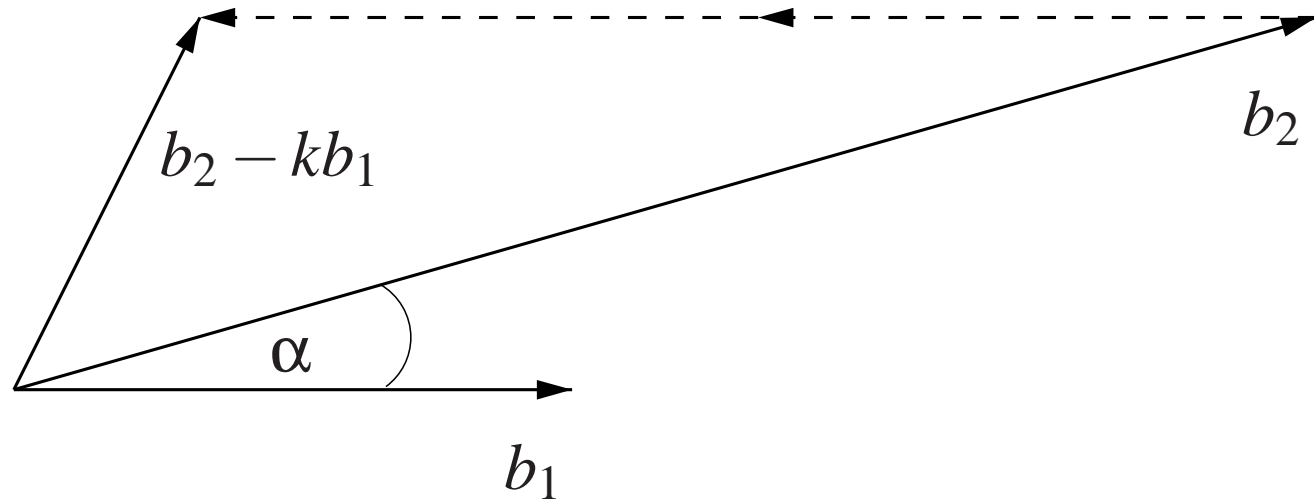
## Reduced bases

Lattice basis  $(b_1, b_2)$  is **reduced** if enclosed angle in range  $90^\circ \pm 30^\circ$



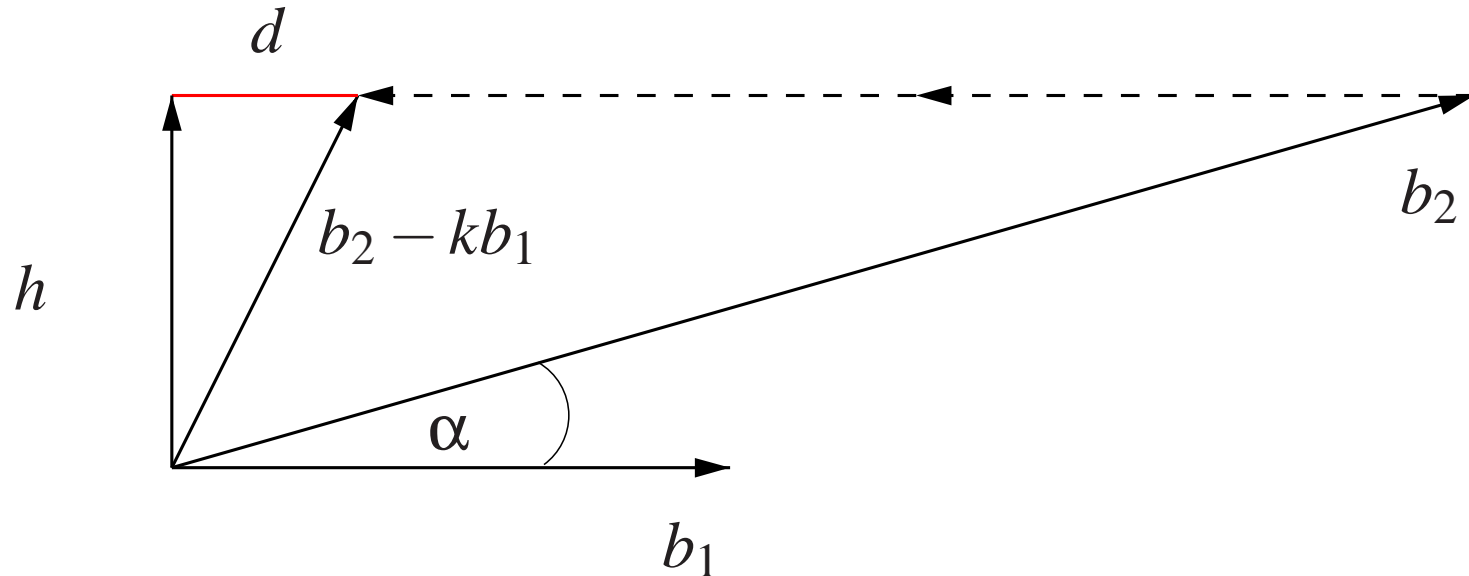
## Reduced bases

Lattice basis  $(b_1, b_2)$  is **reduced** if enclosed angle in range  $90^\circ \pm 30^\circ$



# Reduced bases

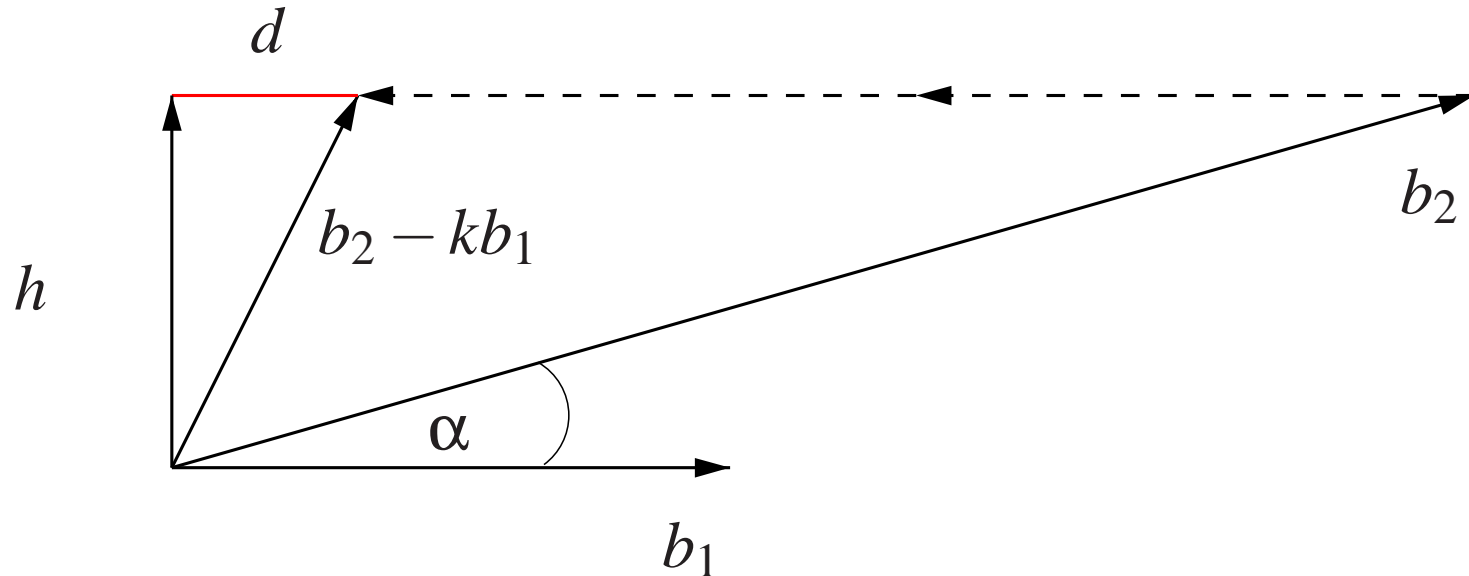
Lattice basis  $(b_1, b_2)$  is **reduced** if enclosed angle in range  $90^\circ \pm 30^\circ$



$$\|d\| \leq 1/2 \|b_1\| \text{ and } \|h\| = \sin \alpha \|b_2\| \quad \text{suppose } \alpha < 60^\circ$$

## Reduced bases

Lattice basis  $(b_1, b_2)$  is **reduced** if enclosed angle in range  $90^\circ \pm 30^\circ$

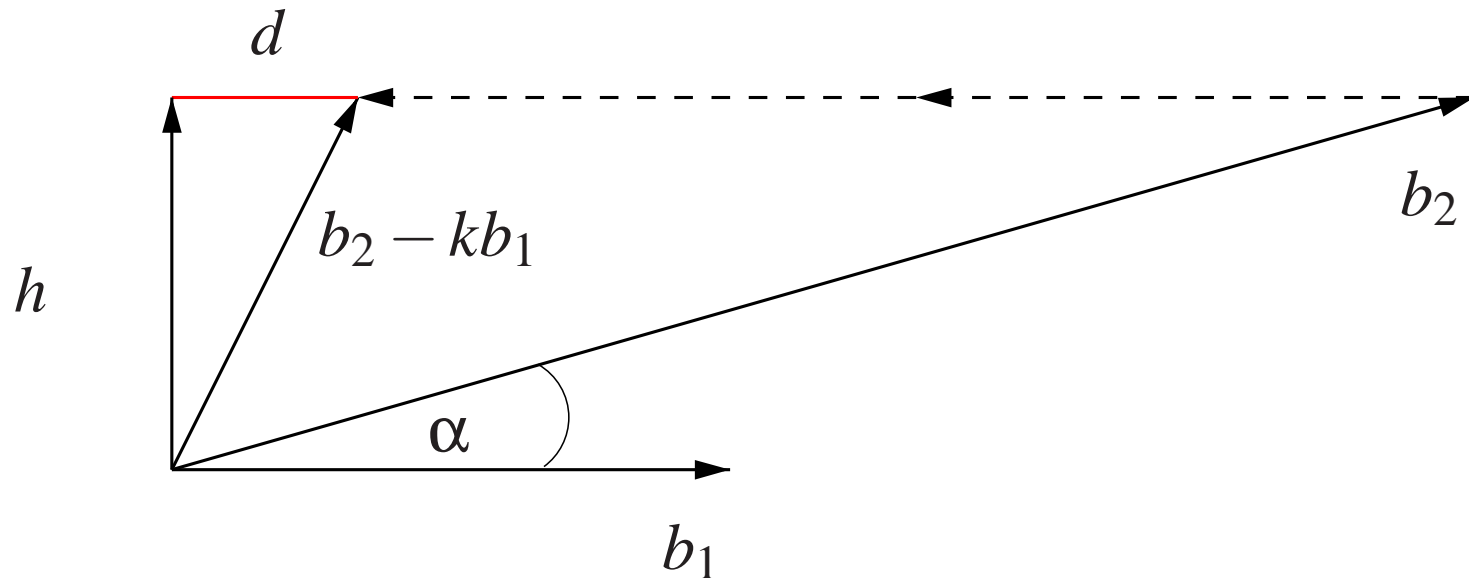


$$\|d\| \leq 1/2 \|b_1\| \text{ and } \|h\| = \sin \alpha \|b_2\| \quad \text{suppose } \alpha < 60^\circ$$

$$\begin{aligned} \|b_2 - kb_1\|^2 &= \|h\|^2 + \|d\|^2 = 1/4 \|b_1\|^2 + (\sin \alpha)^2 \|b_2\|^2 \\ &\leq (1/4 + (\sin \alpha)^2) \|b_2\|^2 \end{aligned}$$

## Reduced bases

Lattice basis  $(b_1, b_2)$  is **reduced** if enclosed angle in range  $90^\circ \pm 30^\circ$



$$\|d\| \leq 1/2 \|b_1\| \text{ and } \|h\| = \sin \alpha \|b_2\| \quad \text{suppose } \alpha < 60^\circ$$

$$\begin{aligned} \|b_2 - kb_1\|^2 &= \|h\|^2 + \|d\|^2 = 1/4 \|b_1\|^2 + (\sin \alpha)^2 \|b_2\|^2 \\ &\leq (1/4 + (\sin \alpha)^2) \|b_2\|^2 < \|b_2\|^2 \end{aligned}$$

## Almost reduced basis

**Proposition.** *There exists  $O(M(n) \log n)$  time algorithm that computes basis  $B$  of  $\Lambda$  defined by  $A \in \mathbf{Z}^{2 \times 2}$ , with property that the first column of  $B$  is shortest vector w.r.t.  $\ell_\infty$ -norm.*



## Almost reduced basis

**Proposition.** *There exists  $O(M(n) \log n)$  time algorithm that computes basis  $B$  of  $\Lambda$  defined by  $A \in \mathbf{Z}^{2 \times 2}$ , with property that the first column of  $B$  is shortest vector w.r.t.  $\ell_\infty$ -norm.*

Analysis of Gaussian algorithm (Lagarias 1980) reveals this basis  
almost reduced

## Almost reduced basis

**Proposition.** *There exists  $O(M(n) \log n)$  time algorithm that computes basis  $B$  of  $\Lambda$  defined by  $A \in \mathbf{Z}^{2 \times 2}$ , with property that the first column of  $B$  is shortest vector w.r.t.  $\ell_\infty$ -norm.*

Analysis of Gaussian algorithm (Lagarias 1980) reveals this basis  
almost reduced

**Corollary.** *There exists  $O(M(n) \log n)$  time algorithm that computes reduced basis  $B$  of  $\Lambda$  defined by  $A \in \mathbf{Z}^{2 \times 2}$*

## Summary of results

- Shortest vector corresponds to best approximation of a rational defined by the lattice
- Shortest vectors can be found with the euclidean algorithm
- Fast basis reduction can be solely based on Schönhage's (1971) result and reduction algorithm of Gauss (1801)

## Summary of results

- Shortest vector corresponds to best approximation of a rational defined by the lattice
- **Shortest vectors can be found with the euclidean algorithm**
- Fast basis reduction can be solely based on Schönhage's (1971) result and reduction algorithm of Gauss (1801)

## Summary of results

- Shortest vector corresponds to best approximation of a rational defined by the lattice
- Shortest vectors can be found with the euclidean algorithm
- Fast basis reduction can be solely based on Schönhage's (1971) result and reduction algorithm of Gauss (1801)

# Open Problem

What about shortest vector problem in arbitrary **fixed** dimension?

## Open Problem

What about shortest vector problem in arbitrary **fixed** dimension?

Fastest known algorithm of Kannan (1987) runs in time  $O(M(n)n)$

## Open Problem

What about shortest vector problem in arbitrary **fixed** dimension?

Fastest known algorithm of Kannan (1987) runs in time  $O(M(n)n)$

**Challenge:**

Prove that shortest vector problem in arbitrary fixed dimension can be solved in time  $O(M(n) \log n)$