

# The unique-SVP World

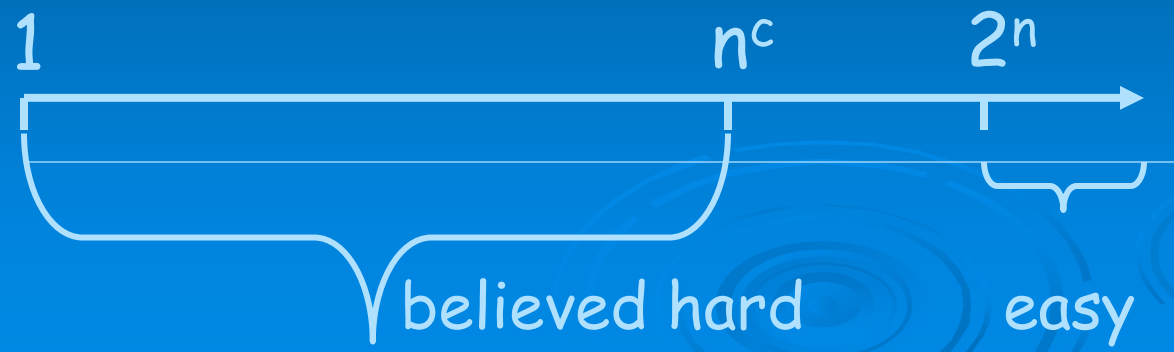
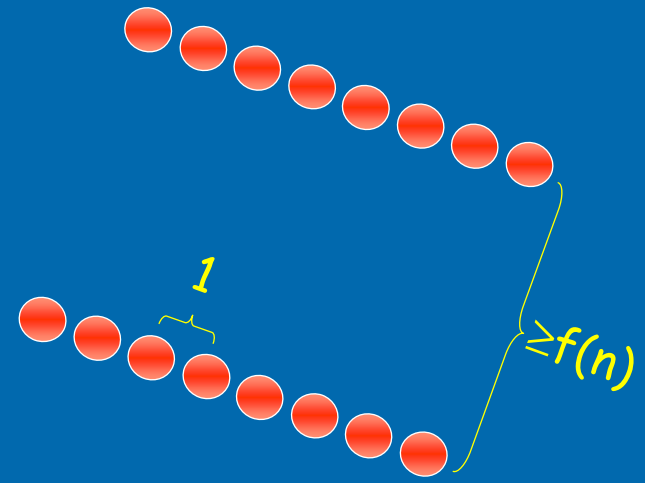
Shai Halevi, IBM, July 2009

1. Ajtai-Dwork'97/07, Regev'03
  - PKE from worst-case uSVP
2. Lyubashvsky-Micciancio'09
  - Relations between worst-case uSVP, BDD, GapSVP

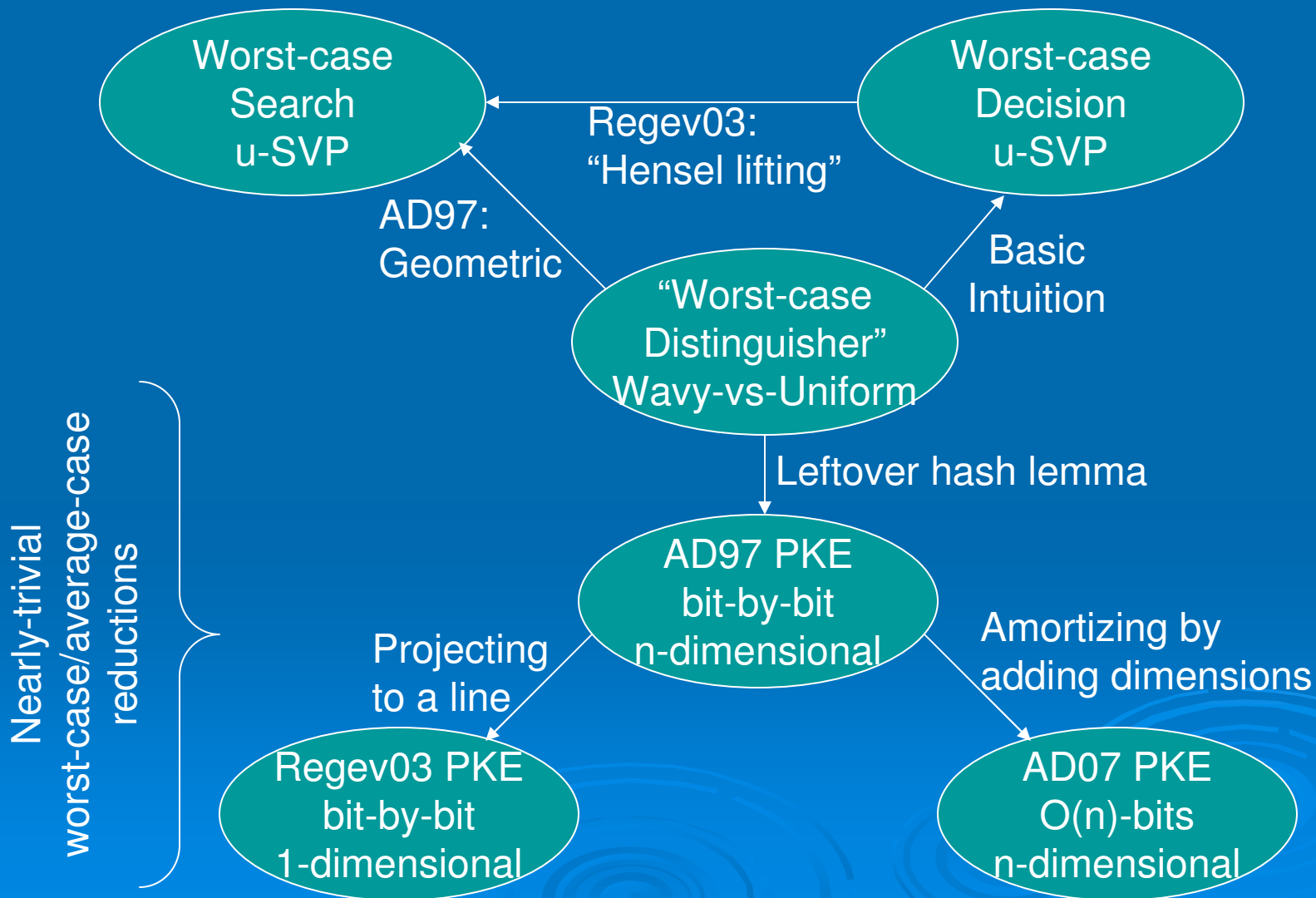
Many slides stolen from Oded Regev, denoted by ®

# $f(n)$ -unique-SVP

- Promise: the shortest vector  $u$  is shorter by a factor of  $f(n)$
- Algorithm for  $2^n$ -unique SVP [LLL82, Schnorr87]
- Believed to be hard for any polynomial  $n^c$



# Ajtai-Dwork & Regev'03 PKEs



# n-dimensional distributions <sup>®</sup>

- Distinguish between the distributions:



Wavy

(In a random direction)

?

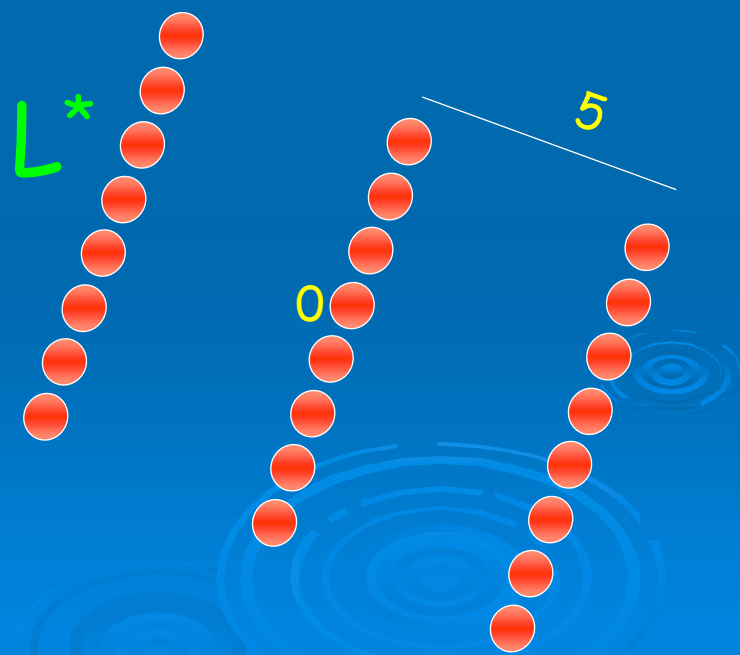
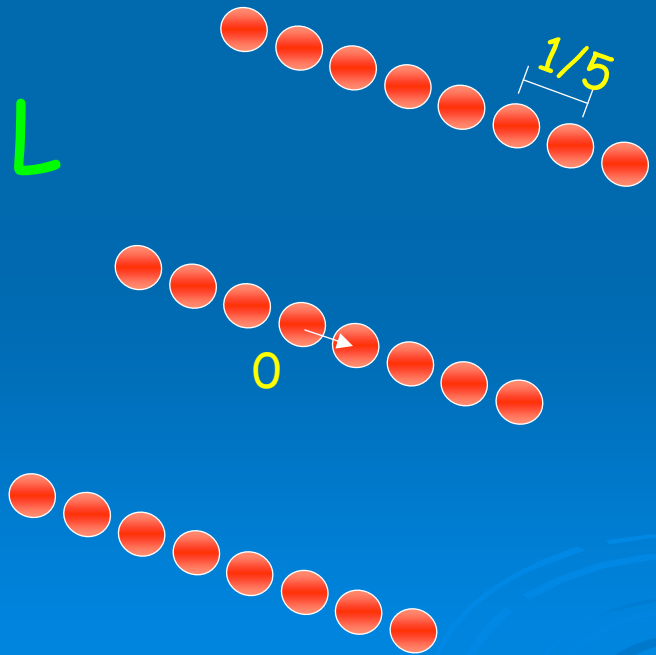


Uniform

# Dual Lattice

➤ Given a lattice  $L$ , the dual lattice is

$$L^* = \{ x \mid \text{for all } y \in L, \langle x, y \rangle \in \mathbb{Z} \}$$



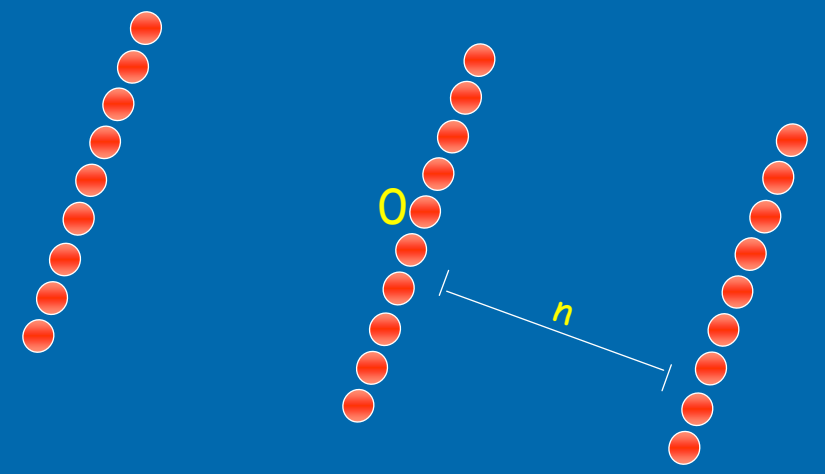
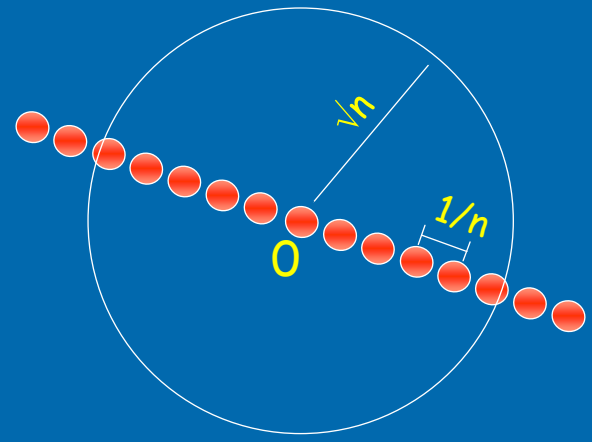
# $L^*$ - the dual of $L$

®

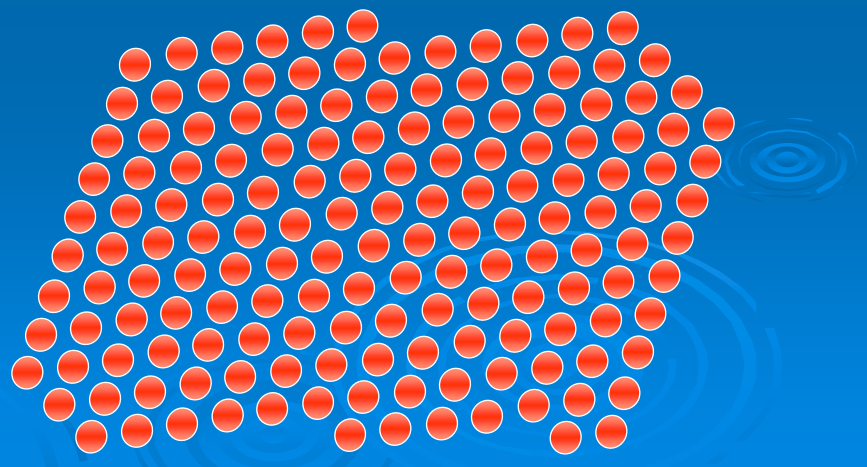
$L$

$L^*$

Case 1



Case 2



# Reduction

- Input: a basis  $B^*$  for  $L^*$
- Produce a distribution that is:
  - Wavy if  $L$  has unique shortest vector ( $|u| \leq 1/n$ )
  - Uniform (on  $P(B^*)$ ) if  $\lambda_1(L) > \sqrt{n}$
- Choose a point from a Gaussian of radius  $\sqrt{n}$ , and reduce mod  $P(B^*)$ 
  - Conceptually, a "random  $L^*$  point" with a  $\text{Gaussian}(\sqrt{n})$  perturbation

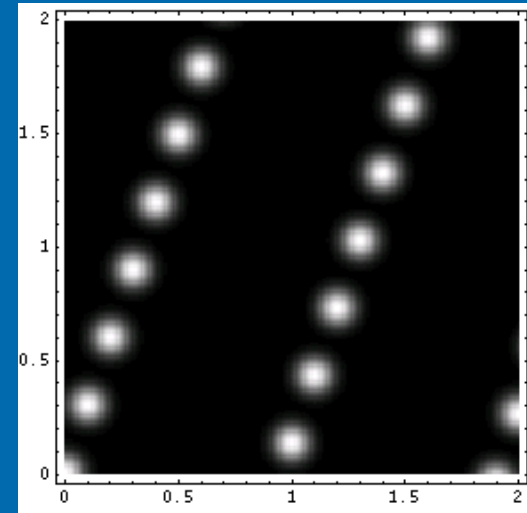
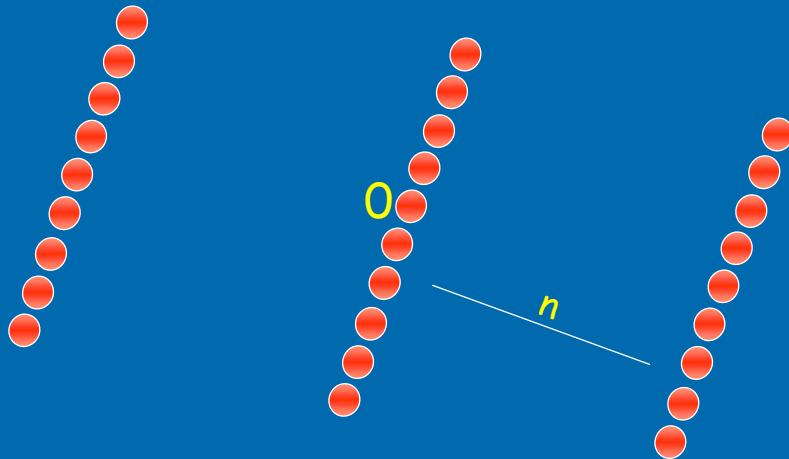
# Creating the Distribution

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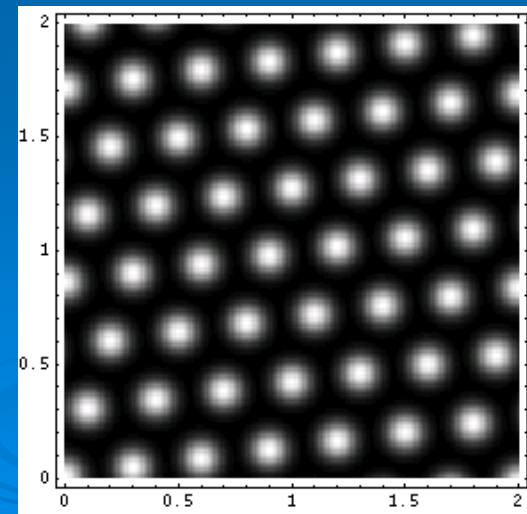
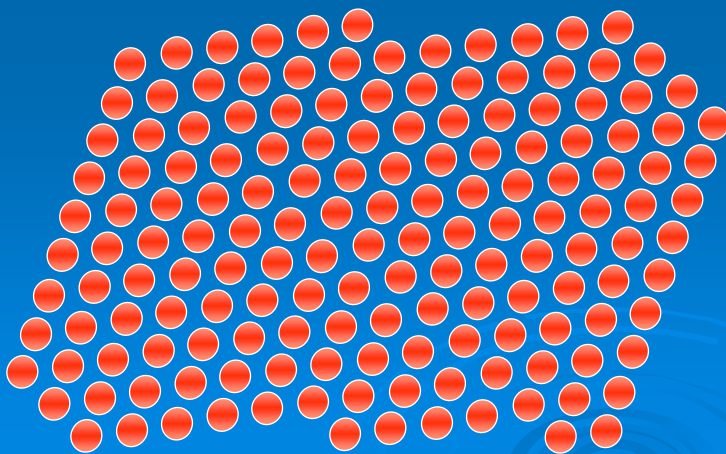
$L^*$

$L^* + \text{perturb}$

Case 1



Case 2





# Analyzing the Distribution <sup>®</sup>

➤ Theorem: (using [Banaszczyk'93])

The distribution obtained above depends only on the points in  $L$  of distance  $\sqrt{n}$  from the origin (up to an exponentially small error)

➤ Therefore,

Case 1: Determined by multiples of  $u \rightarrow$

wavy on hyperplanes orthogonal to  $u$

Case 2: Determined by the origin  $\rightarrow$

uniform

# Proof of Theorem ®

- For a set  $A$  in  $\mathbb{R}^n$ , define:

$$\rho(A) = \sum_{x \in A} e^{-\pi \|x\|^2}$$

- Poisson Summation Formula implies:

$$\forall y \in P(L^*), \rho(y - L^*) = d(L) \cdot \sum_{x \in L} e^{2\pi i \langle x, y \rangle} \rho(\{x\})$$

- Banaszczyk's theorem:

For any lattice  $L$ ,

$$\rho(L - \sqrt{n}B_n) < 2^{-\Omega(n)} \rho(L \cap \sqrt{n}B_n)$$

# Proof of Theorem (cont.)

®

- In Case 2, the distribution obtained is very close to uniform:

$$\begin{aligned} \forall y \in P(L^*), \rho(y - L^*) &= d(L) \cdot \sum_{x \in L} e^{2\pi i \langle x, y \rangle} \rho(\{x\}) = \\ &d(L) \cdot \left( 1 + \sum_{x \in L - \{0\}} e^{2\pi i \langle x, y \rangle} \rho(\{x\}) \right) \approx d(L) \end{aligned}$$

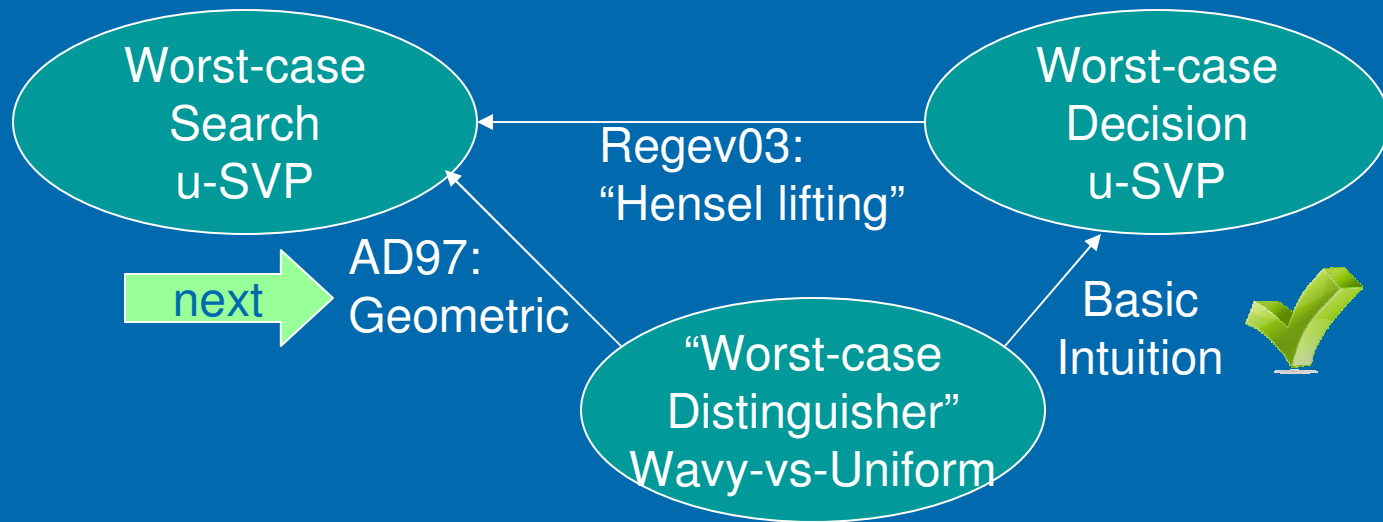
- Because:

$$\left| \sum_{x \in L - \{0\}} e^{2\pi i \langle x, y \rangle} \rho(\{x\}) \right| < \sum_{x \in L - \{0\}} \rho(\{x\}) =$$

$$\rho(L - \{0\}) = \rho(L - \sqrt{n}B_n) <$$

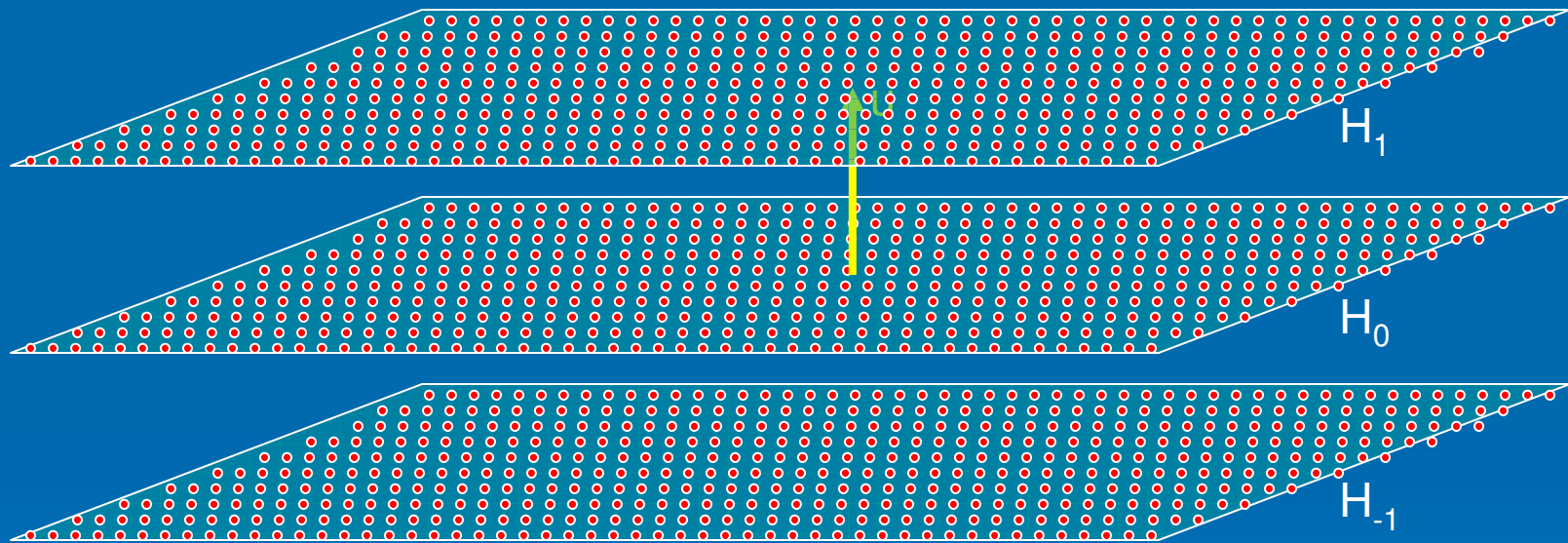
$$2^{-\Omega(n)} \rho(L \cap \sqrt{n}B_n) = 2^{-\Omega(n)}$$

# Ajtai-Dwork & Regev'03 PKEs



# Distinguish $\rightarrow$ Search, AD97

- Reminder:  $L^*$  lives in hyperplanes



- We want to identify  $u$ 
  - Using an oracle that distinguishes wavy distributions from uniform in  $P(B^*)$

# The plan

1. Use the oracle to distinguish points close to  $H_0$  from points close to  $H_{\pm 1}$
2. Then grow very long vectors that are rather close to  $H_0$
3. This gives a very good approximation for  $u$ , then we use it to find  $u$  exactly

# Distinguishing $H_0$ from $H_{\pm 1}$

Input: basis  $B^*$  for  $L^*$ ,  $\sim$ length of  $u$ , point  $x$

- And access to wavy/uniform distinguisher

Decision: **Is  $x$   $1/\text{poly}(n)$  close to  $H_0$  or to  $H_{\pm 1}$ ?**

- Choose  $y$  from a wavy distribution near  $L^*$ 
  - $y = \text{Gaussian}(\sigma)^*$  with  $\sigma < 1/2|u|$
- Pick  $\alpha \in_{\mathbb{R}} [0, 1]$ , set  $z = \alpha x + y \bmod P(B^*)$
- Ask oracle if  $z$  is drawn from wavy or uniform distribution

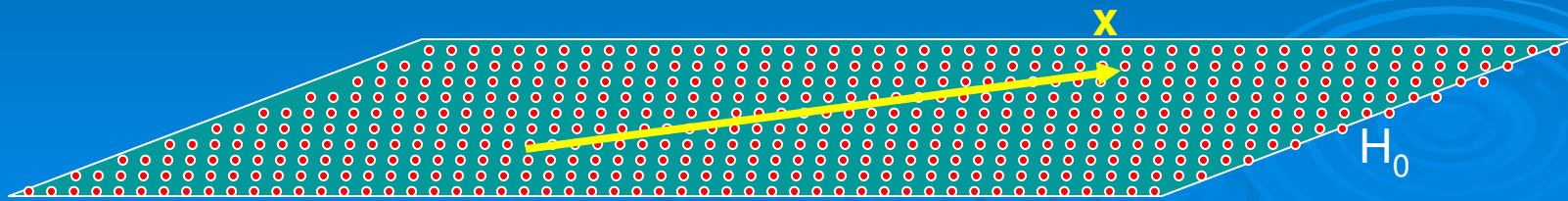
\*  $\text{Gaussian}(\sigma)$ : variance  $\sigma^2$  in each coordinate

# Distinguishing $H_0$ from $H_{\pm 1}$ (cont.)

Case 1:  $x$  close to  $H_0$

➤  $\alpha x$  also close to  $H_0$

➤  $\alpha x + y \bmod P(B^*)$  close to  $L^*$ , wavy

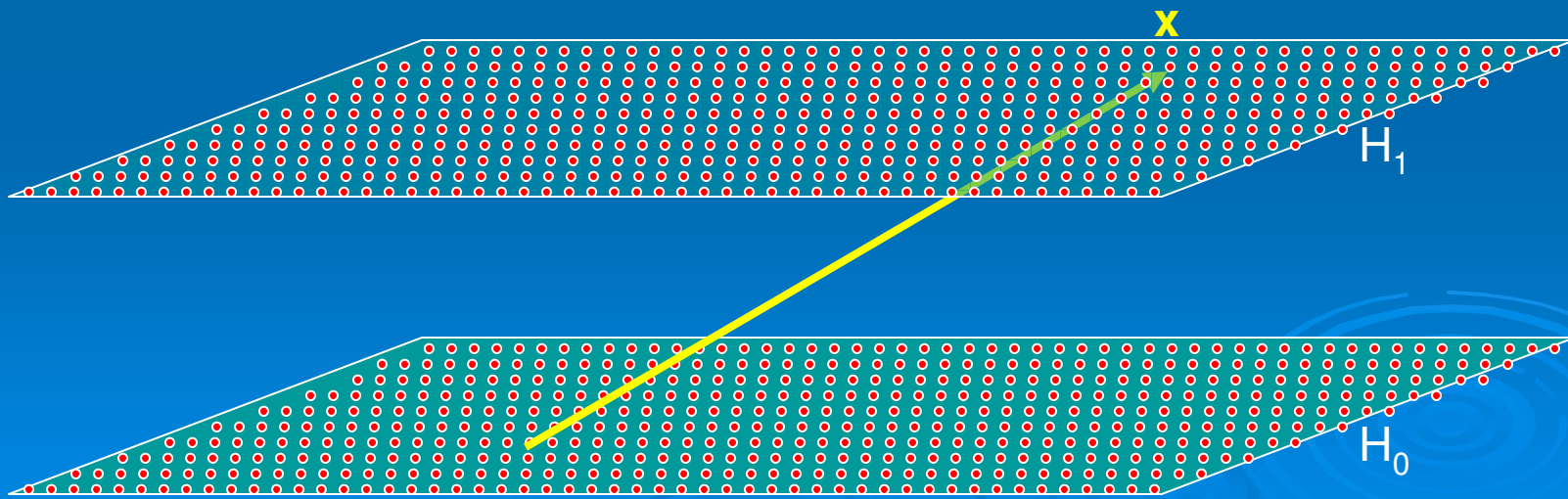




# Distinguishing $H_0$ from $H_{\pm 1}$ (cont.)

Case 2:  $x$  close to  $H_{\pm 1}$

- $\alpha x$  “in the middle” between  $H_0$  and  $H_{\pm 1}$ 
  - Nearly uniform component in the  $u$  direction
- $\alpha x + y \bmod P(B^*)$  nearly uniform in  $P(B^*)$

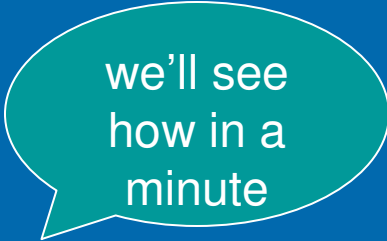


# Distinguishing $H_0$ from $H_{\pm 1}$ (cont.)

- Repeat  $\text{poly}(n)$  times, take majority
  - Boost the advantage to near-certainty
- Below we assume a “perfect distinguisher”
  - Close to  $H_0 \rightarrow$  always says NO
  - Close to  $H_{\pm 1} \rightarrow$  always says YES
  - Otherwise, there are no guarantees
    - Except halting in polynomial time

# Growing Large Vectors

- Start from some  $x_0$  between  $H_{-1}$  and  $H_{+1}$ 
  - e.g. a random vector of length  $1/|u|$
- In each step, choose  $x_i$  s.t.
  - $|x_i| \sim 2|x_{i-1}|$
  - $x_i$  is somewhere between  $H_{-1}$  and  $H_{+1}$
- Keep going for  $\text{poly}(n)$  steps
- Result is  $x^*$  between  $H_{\pm 1}$  with  $|x^*| = N/|u|$ 
  - Very large  $N$ , e.g.,  $N=2^{n^2}$

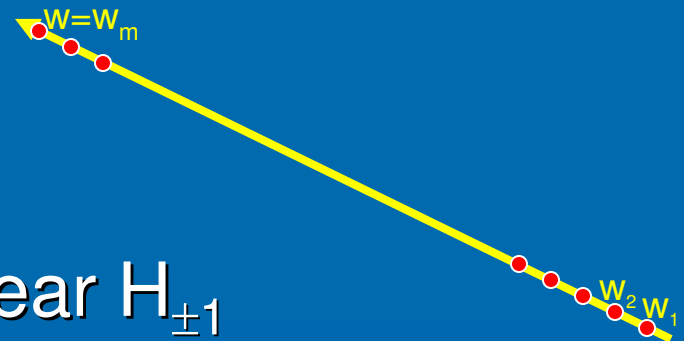


we'll see  
how in a  
minute

# From $x_{i-1}$ to $x_i$

Try poly(n) many candidates:

- Candidate  $w = 2x_{i-1} + \text{Gaussian}(1/|u|)$
- For  $j = 1, \dots, m = \text{poly}(n)$ 
  - $w_j = j/m \cdot w$
  - Check if  $w_j$  is near  $H_0$  or near  $H_{\pm 1}$
- If none of the  $w_j$ 's is near  $H_{\pm 1}$  then accept  $w$  and set  $x_i = w$
- Else try another candidate

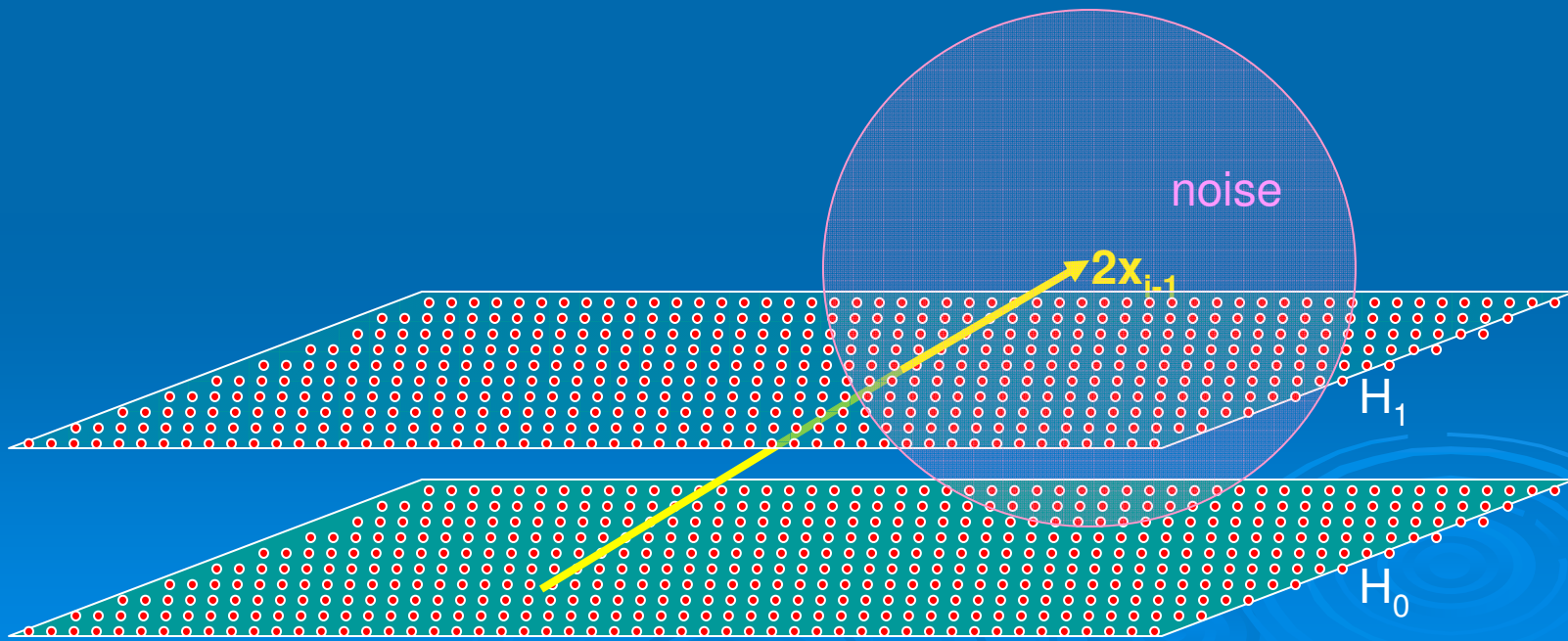


# From $x_{i-1}$ to $x_i$ : Analysis

- $x_{i-1}$  between  $H_{\pm 1}$   $\rightarrow$   $w$  is between  $H_{\pm n}$ 
  - Except with exponentially small probability
- $w$  is NOT between  $H_{\pm 1}$   $\rightarrow$  some  $w_j$  near  $H_{\pm 1}$ 
  - So  $w$  will be rejected
- So if we make progress, we know that we are on the right track

# From $x_{i-1}$ to $x_i$ : Analysis (cont.)

- With probability  $1/\text{poly}(n)$ ,  $w$  is close to  $H_0$ 
  - The component in the  $u$  direction is Gaussian with mean  $< 2/|u|$  and variance  $1/|u|^2$



# From $x_{i-1}$ to $x_i$ : Analysis (cont.)

- With probability  $1/\text{poly}$ ,  $w$  is close to  $H_0$ 
  - The component in the  $u$  direction is Gaussian with mean  $< 2/|u|$  and standard deviation  $1/|u|$
- $w$  is close to  $H_0$ , all  $w_j$ 's are close to  $H_0$ 
  - So  $w$  will be accepted
- After polynomially many candidates, we will make progress whp

# Finding $u$

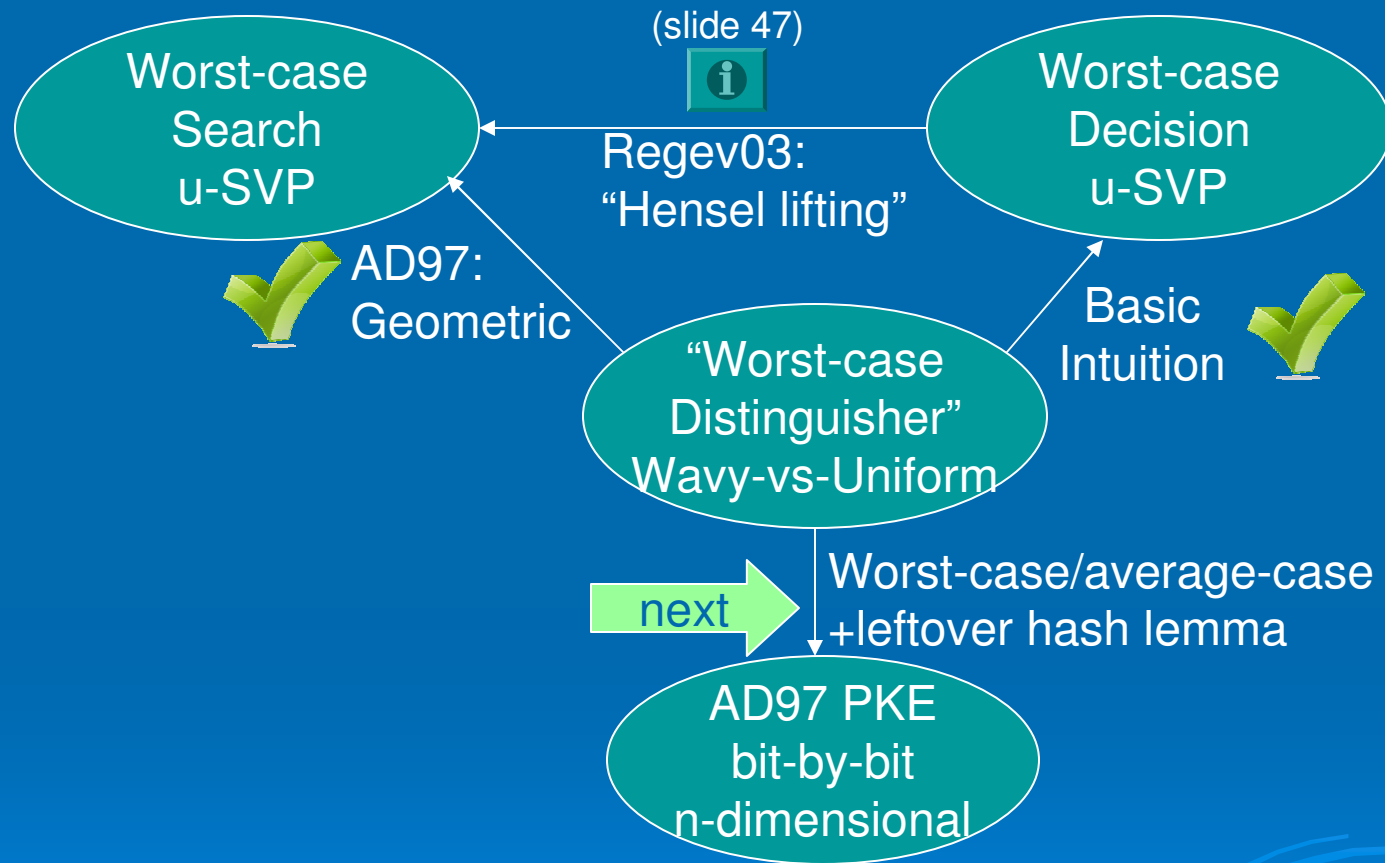
- Find  $n-1$   $x^*$ 's
  - $x^*_{t+1}$  is chosen orthogonal to  $x^*_1, \dots, x^*_t$
  - By choosing the Gaussians in that subspace
- Compute  $u' \perp \{x^*_1, \dots, x^*_{n-1}\}$ , with  $|u'|=1$ 
  - $u'$  is exponentially close to  $u/|u|$ 
    - $u/|u| = (u' + e)$ ,  $|e|=1/N$
    - Can make  $N \gg 2^n$  (e.g.,  $N=2^{n^2}$ )
- Diophantine approximation to solve for  $u$



(slide 71)



# Ajtai-Dwork & Regev'03 PKEs



# Average-case Distinguisher

- Intuition: lattice only matters via the direction of  $u$
- Security parameter  $n$ , another parameter  $N$
- A random  $u$  in  $n$ -dim. unit sphere defines  $\mathcal{D}_u(N)$ 
  - $\chi = \text{discret-Gaussian}(N)$  in one dimension
    - Defines a vector  $x = \chi \cdot u / \langle u, u \rangle$ , namely  $x \parallel u$  and  $\langle x, u \rangle = \chi$
  - $y = \text{Gaussian}(N)$  in the other  $n-1$  dimensions
  - $e = \text{Gaussian}(n^{-4})$  in all  $n$  dimensions
  - Output  $x+y+e$

# Worst-case/average-case (cont.)

Thm: Distinguishing  $\mathcal{D}_u(N)$  from Uniform

→ Distinguishing  $\text{Wavy}_{B^*}$  from  $\text{Uniform}_{B^*}$  for all  $B^*$

- When you know  $\lambda_1(L(B))$  upto  $(1+1/\text{poly}(n))$ -factor
- For parameter  $N = 2^{\Omega(N)}$

Pf: Given  $B^*$ , scale it s.t.  $\lambda_1(L(B)) \in [1, 1+1/\text{poly}]$

➤ Also apply random rotation

➤ Given samples  $x$  (from  $\text{Uniform}_{B^*}$  /  $\text{Wavy}_{B^*}$ )

- Sample  $y = \text{discrete-Gaussian}_{B^*}(N)$ 
  - Can do this for large enough  $N$

- Output  $z = x + y$

➤ “Clearly”  $z$  is close to  $\mathcal{G}(N) / \mathcal{D}_u(N)$  respectively

# The AD97 Cryptosystem

- Secret key: a random  $u \in$  unit sphere
- Public key:  $n+m+1$  vectors ( $m=8n \log n$ )
  - $b_1, \dots, b_n \leftarrow \mathcal{D}_u(2^n)$ ,  $v_0, v_1, \dots, v_m \leftarrow \mathcal{D}_u(n2^n)$ 
    - So  $\langle b_i, u \rangle$ ,  $\langle v_i, u \rangle \sim$  integer
    - We insist on  $\langle v_0, u \rangle \sim$  odd integer
- Will use  $P(b_1, \dots, b_n)$  for encryption
  - Need  $P(b_1, \dots, b_n)$  with “width”  $> 2^n/n$

# The AD97 Cryptosystem (cont.)

## Encryption( $\sigma$ ):

- $c' \leftarrow \text{random-subset-sum}(v_1, \dots, v_m) + \sigma v_0/2$
- output  $c = (c' + \text{Gaussian}(n^4)) \bmod P(B)$

## Decryption( $c$ ):

- If  $\langle u, c \rangle$  is closer than  $1/4$  to integer say 0, else say 1

Correctness due to  $\langle b_i, u \rangle, \langle v_j, u \rangle \sim \text{integer}$

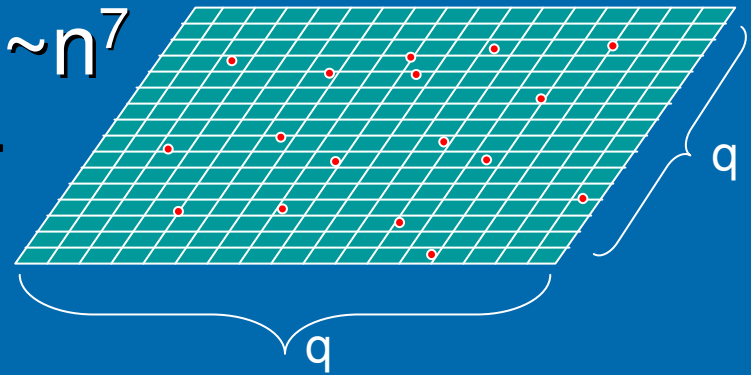
- and width of  $P(B)$

# AD97 Security

- The  $b_i$ 's,  $v_i$ 's chosen from  $\mathcal{D}_u$  (something)
- By hardness assumption, can't distinguish from  $\mathcal{G}_u$  (something)
- Claim: if they were from  $\mathcal{G}_u$  (something),  $c$  would have no information on the bit  $\sigma$ 
  - Proven by leftover hash lemma + smoothing
- Note:  $v_i$ 's has variance  $n^2$  larger than  $b_i$ 's
  - In the  $\mathcal{G}_u$  case  $v_i \bmod P(B)$  is nearly uniform

# AD97 Security (cont.)

- Partition  $P(B)$  to  $q^n$  cells,  $q \sim n^7$
- For each point  $v_i$ , consider the cell where it lies
  - $r_i$  is the corner of that cell
- $\sum_S v_i \bmod P(B) = \sum_S r_i \bmod P(B) + n^{-5}$  “error”
  - $S$  is our random subset
- $\sum_S r_i \bmod P(B)$  is a nearly-random cell
  - We’ll show this using leftover hash
- The Gaussian( $n^{-4}$ ) in  $c$  drowns the error term



# Leftover Hashing

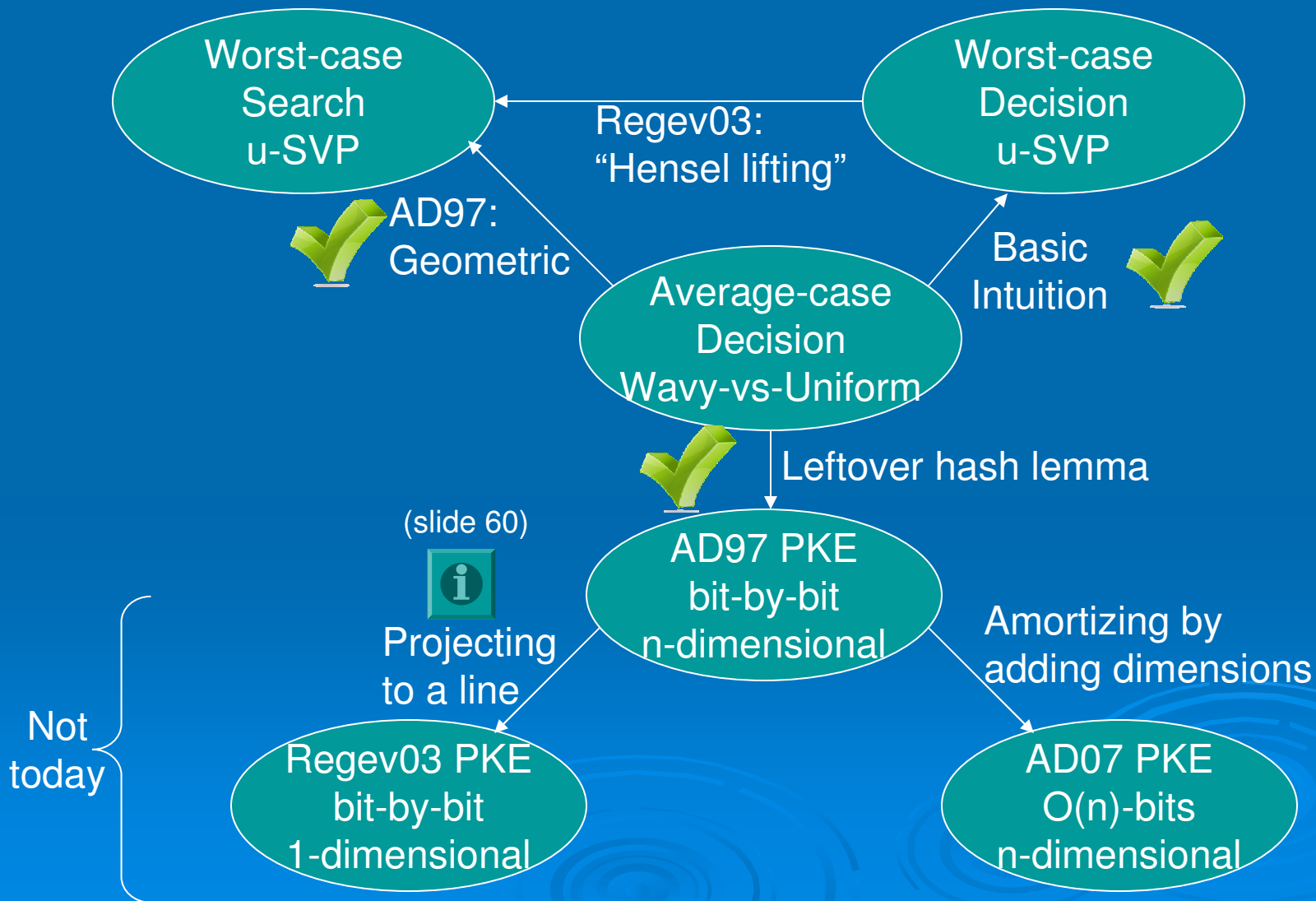
- Consider hash function  $H_R: \{0,1\}^m \rightarrow [q]^n$ 
  - The key is  $R = [r_1, \dots, r_m] \in [q]^{n \times m}$
  - The input is a bit vector  $b = [\sigma_1, \dots, \sigma_m]^T \in \{0,1\}^m$
- $H_R(b) = Rb \pmod q$
- $H$  is “pairwise independent” (well, almost..)
  - Yay, let’s use the leftover hash lemma
- $\langle R, H_R(b) \rangle, \langle R, \mathcal{U} \rangle$  statistically close
  - For random  $R \in [q]^{n \times m}$ ,  $b \in \{0,1\}^m$ ,  $\mathcal{U} \in [q]^n$
  - Assuming  $m \gg n \log q$



# AD97 Security (cont.)

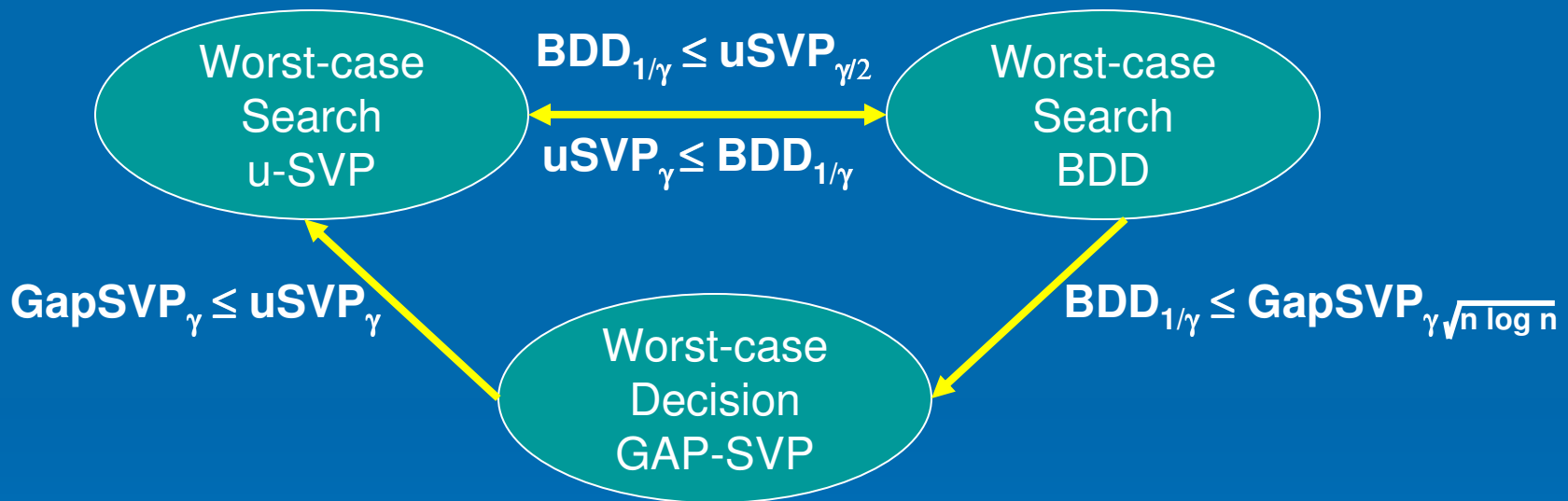
- We proved  $\sum_S r_i \bmod P(B)$  is nearly-random
- Recall:
  - $c_0 = \sum_S r_i + \text{error}(n^{-5}) + \text{Gaussian}(n^{-4}) \bmod P(B)$
- For any  $x$  and error  $e$ ,  $|e| \sim n^{-5}$ , the distr.  $x+e+\text{Gaussian}(n^{-5})$ ,  $x+\text{Gaussian}(n^{-4})$  are statistically close
- So  $c_0 \sim \sum_S r_i + \text{Gaussian}(n^{-3}) \bmod P(B)$ 
  - Which is close to uniform in  $P(B)$
  - Also  $c_1 = c_0 + v_0/2 \bmod P(B)$  close to uniform

# Ajtai-Dwork & Regev'03 PKEs



# u-SVP vs. BDD vs. GAP-SVP

- Lyubashevsky-Micciancio, CRYPTO 2009



- Good old-fashion worst-case reductions
  - Mostly Cook reductions (one Karp reduction)

# Reminder: uSVP and BDD

**uSVP $_{\gamma}$** :  $\gamma$ -unique shortest vector problem

- Input: a basis  $B = (b_1, \dots, b_n)$
- Promise:  $\lambda_1(L(B)) < \gamma \lambda_2(L(B))$
- Task: find shortest nonzero vector in  $L(B)$

**BDD $_{1/\gamma}$** :  $1/\gamma$ -bounded distance decoding

- Input: a basis  $B = (b_1, \dots, b_n)$ , a point  $t$
- Promise:  $\text{dist}(t, L(B)) < \lambda_1(L(B)) / \gamma$
- Task: find closest vector to  $t$  in  $L(B)$

$$\text{BDD}_{1/\gamma} \leq \text{uSVP}_{\gamma/2}$$

- Input: a basis  $B = (b_1, \dots, b_n)$ , a point  $t$ 
  - Assume that we know  $\mu = \text{dist}(t, L(B))$

- Let  $B' = \begin{pmatrix} b_1 & \dots & b_n & t \\ 0 & & 0 & \mu \end{pmatrix}$

Can get by with a good approximation for  $\mu$

- Let  $v \in L(B)$  be the closest to  $t$ ,  $|t-v|=\mu$
  - Will show that the vector  $[(t-v) \ \mu]^T$  is the  $\gamma/2$ -unique shortest vector in  $L(B')$
  - So  $\text{uSVP}_{\gamma/2}(B')$  will return it
- The size of  $v' = [(t-v) \ \mu]^T$  is  $(\mu^2 + \mu^2)^{1/2} = \sqrt{2} \times \mu$

# BDD $_{1/\gamma} \leq \text{uSVP}_{\gamma/2}$ (cont.)

- Every  $w' \in L(B')$  looks like  $w' = [\beta t - w \ \beta \mu]^T$ 
  - For some integer  $\beta$  and some  $w \in L(B)$
  - Write  $\beta t - w = (\beta v - w) - \beta(v - t)$
  - $\beta v - w \in L(B)$ , nonzero if  $w'$  isn't a multiple of  $v'$
  - So  $|\beta v - w| \geq \lambda_1$ , also recall  $|v - t| = \mu \leq \lambda_1/\gamma$
  - ➔  $|\beta t - w| \geq |\beta v - w| - \beta|v - t| \geq \lambda_1 - \beta\mu$
  - ➔  $|w'|^2 \geq (\lambda_1 - \beta\mu)^2 + (\beta\mu)^2 \geq \inf_{\beta \in \mathbb{R}} [(\lambda_1 - \beta\mu)^2 + (\beta\mu)^2]$   
 $= (\lambda_1)^2/2 \geq (\gamma\mu)^2/2$
- So for any  $w' \in L(B')$ , not a multiple of  $v'$ , we have  $|w'| \geq \mu\gamma/\sqrt{2} = |v'| \times \gamma/2$

$$\text{uSVP}_\gamma \leq \text{BDD}_{1/\gamma}$$

- Input: a basis  $B = (b_1, b_2, \dots, b_n)$ 
  - Let  $\rho$  be a prime,  $\rho \geq \gamma$
- For  $i=1, 2, \dots, n$ ,  $j=1, 2, \dots, \rho-1$ 
  - $B_i = (b_1, b_2, \dots, \rho \times b_i, \dots, b_n)$ ,  $t_{ij} = j \times b_i$
  - Let  $v_{ij} = \text{BDD}_{1/\gamma}(B_i, T_{ij})$ ,  $w_{ij} = v_{ij} - t_{ij}$
- Output the smallest nonzero  $w_{ij}$  in  $L(B)$

# $uSVP_\gamma \leq BDD_{1/\gamma}$ (cont.)

- Let  $u$  be shortest nonzero vector in  $L(B)$ 
  - $u = \sum \xi_i b_i$ , at least one  $\xi_i$  isn't divisible by  $\rho$  (otherwise  $u/\rho$  would also be in  $L(B)$ )
  - Let  $j = -\xi_i \bmod \rho$ ,  $j \in \{1, 2, \dots, \rho-1\}$
- We will prove that for these  $i, j$ 
  - $\lambda_1(L(B_i)) > \gamma \lambda_1(L(B))$
  - $\text{dist}(t_{ij}, L(B_i)) \leq \lambda_1(L(B))$



➤ The smallest multiple of  $u$  in  $L(B_i)$  is  $\rho u$

- $|\rho u| = \rho \lambda_1(L(B)) \geq \gamma \lambda_1(L(B))$
- Any other vector in  $L(B_i) \subseteq L(B)$  is longer than  $\gamma \lambda_1(L(B))$  (since  $L(B)$  is  $\gamma$ -unique)

$$\rightarrow \lambda_1(L(B_i)) \geq \gamma \lambda_1(L(B))$$

divisible by  $p$

➤  $t_{ij} + u = j b_i + \sum \xi_m b_m = (j + \xi_i) b_i + \sum_{m \neq i} \xi_m b_m \in L(B_i)$

$$\rightarrow \text{dist}(t_{ij}, L(B_i)) \leq \lambda_1(L(B_i))$$

→  $(B_i, t_{ij})$  satisfies the promise of  $\text{BDD}_{1/\gamma}$

→  $v_{ij} = \text{BDD}_{1/\gamma}(B_i, t_{ij})$  is closest to  $t_{ij}$  in  $L(B_i)$

- $w_{ij} = v_{ij} - t_{ij} \in L(B)$ , since  $t_{ij} \in L(B)$  and  $v_{ij} \in L(B_i) \subseteq L(B)$
- $|w_{ij}| = \lambda_1(L(B))$



# Reminder: GapSVP

- $\text{GapSVP}_\gamma$ : decision version of  $\text{approx}_\gamma\text{-SVP}$ 
  - Input: Basis  $B$ , number  $\delta$
  - Promise: either  $\lambda_1(L(B)) \leq \delta$  or  $\lambda_1(L(B)) > \gamma\delta$
  - Task: decide which is the case
- The reduction  $\text{uSVP}_\gamma \leq \text{GapSVP}_\gamma$  is the same as Regev's Decision-to-Search uSVP reduction



(slide 47)

# GapSVP $_{\gamma\sqrt{n \log n}} \leq$ BDD $_{1/\gamma}$

- Inputs: Basis  $B=(b_1, \dots, b_n)$ , number  $\delta$
- Repeat  $\text{poly}(n)$  times
  - Choose a random  $s_i$  of length  $\leq \delta\sqrt{n \log n}$
  - Set  $t_i = s_i \bmod B$ , run  $v_i = \text{BDD}_{1/\gamma}(B, t_i)$
- Answer YES if  $\exists i$  s.t.  $v \neq t_i - s_i$ , else NO

Need will show:

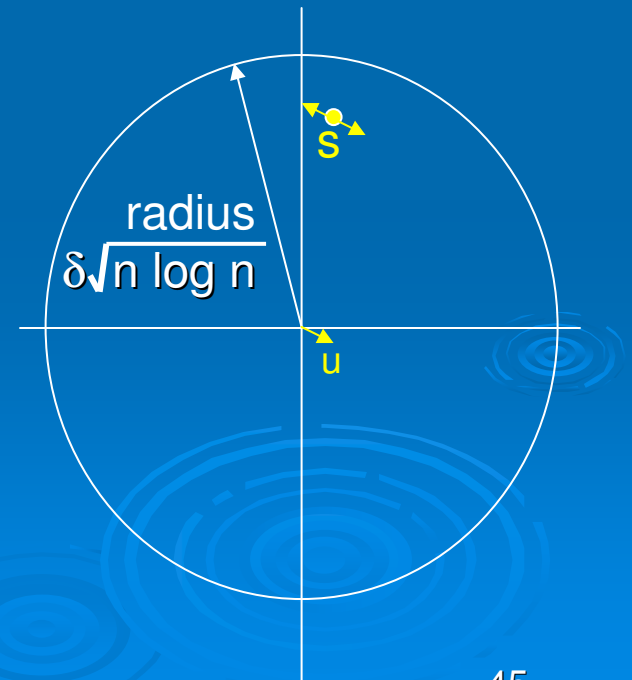
- $\lambda_1(L(B)) > \gamma\delta\sqrt{n \log n} \rightarrow v = t_i - s_i$  always
- $\lambda_1(L(B)) \leq \delta \rightarrow v \neq t_i - s_i$  with probability  $\sim 1/2$

# Case 1: $\lambda_1(L(B)) > \gamma \sqrt{n \log n} \cdot \delta$

- Recall:  $|s_i| \leq \delta \sqrt{n \log n}$ ,  $t_i = s_i \bmod B$ 
  - $t_i$  is  $\leq \delta \sqrt{n \log n}$  away from  $v_i = t_i - s_i \in L(B)$
  - $(B, t_i)$  satisfies the promise of  $\text{BDD}_{1/\gamma}$
  - $\text{BDD}_{1/\gamma}(B, t_i)$  will return some vector in  $L(B)$
- Any other  $L(B)$  point has distance from  $t_i$  at least  $\lambda_1(L(B)) - \delta \sqrt{n \log n} > (\gamma - 1) \delta \sqrt{n \log n}$ 
  - $v_i$  is only answer that  $\text{BDD}_{1/\gamma}(B, t_i)$  can return

## Case 2: $\lambda_1(L(B)) \leq \delta$

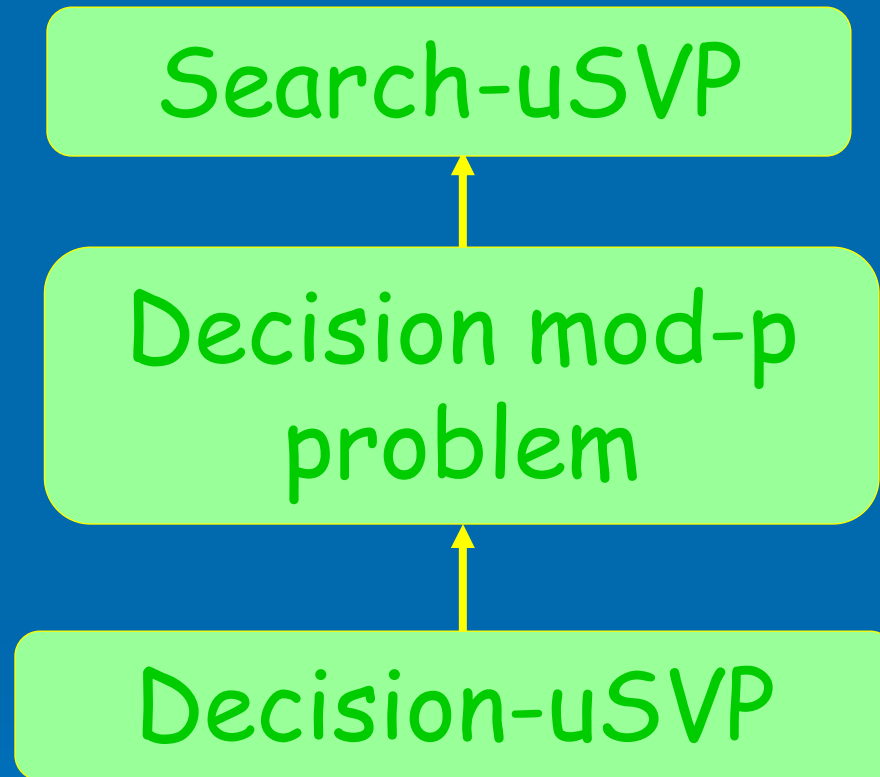
- Let  $u$  be shortest nonzero in  $L(B)$ ,  $|u| = \lambda_1$
- $s_i$  is random in  $\text{Ball}(\delta \sqrt{n \log n})$
- With high probability  $s_i \pm u$  also in ball
  - $t_i = s_i \bmod B$  could just as well be chosen as  $t_i = (s_i + u) \bmod B$
  - Whatever  $\text{BDD}_{1/\gamma}(B, t)$  returns it differs from  $t_i - s_i$  w.p.  $\geq 1/2$



# Backup Slides

1. Regev's Decision-to-Search uSVP
2. Regev's dimension reduction
3. Diophantine Approximation

# uSVP Decision $\rightarrow$ Search



# Reduction from: Decision mod-p



- Given a basis  $(v_1 \dots v_n)$  for  $n^{1.5}$ -unique lattice, and a prime  $p > n^{1.5}$
- Assume the shortest vector is:

$$u = a_1 v_1 + a_2 v_2 + \dots + a_n v_n$$

- Decide whether  $a_1$  is divisible by  $p$



# Reduction to: Decision uSVP



- Given a lattice, distinguish between:
  - Case 1. Shortest vector is of length  $1/n$  and all non-parallel vectors are of length more than  $\sqrt{n}$
  - Case 2. Shortest vector is of length more than  $\sqrt{n}$

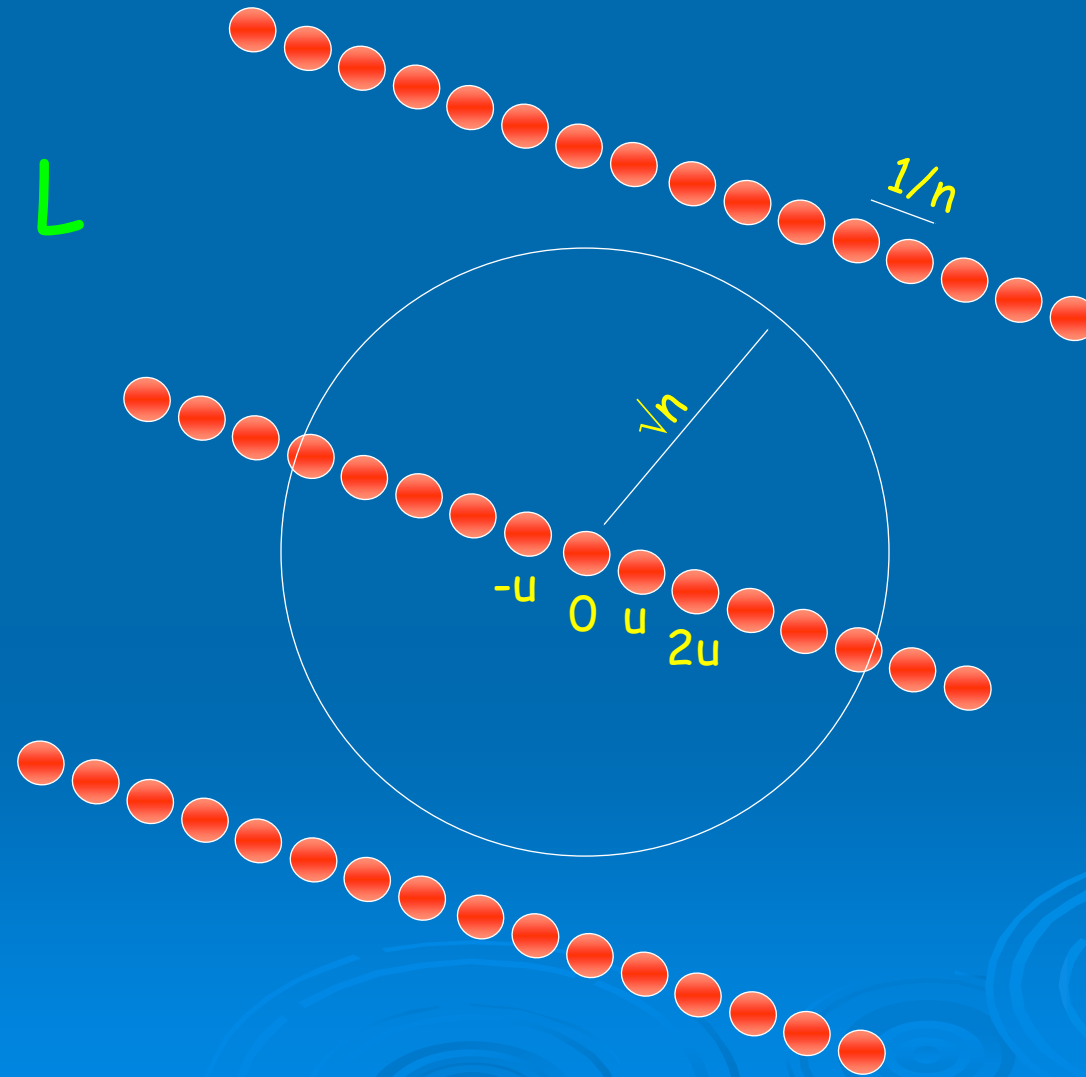
# The reduction



- Input: a basis  $(v_1, \dots, v_n)$  of a  $n^{1.5}$  unique lattice
- Scale the lattice so that the shortest vector is of length  $1/n$
- Replace  $v_1$  by  $pv_1$ . Let  $M$  be the resulting lattice
- If  $p \mid a_1$  then  $M$  has shortest vector  $1/n$  and all non-parallel vectors more than  $\sqrt{n}$
- If  $p \nmid a_1$  then  $M$  has shortest vector more than  $\sqrt{n}$

# The input lattice $L$

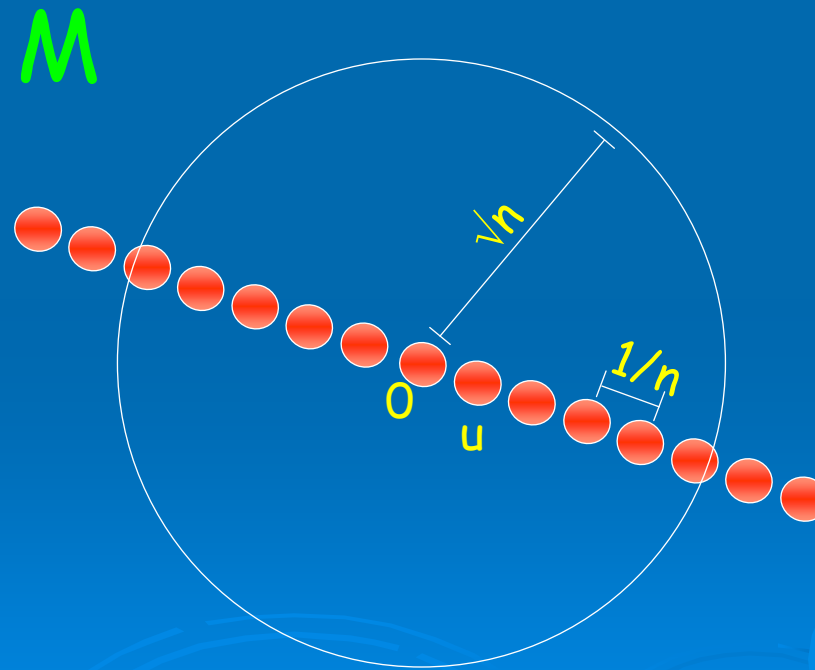
®



# The lattice $M$

®

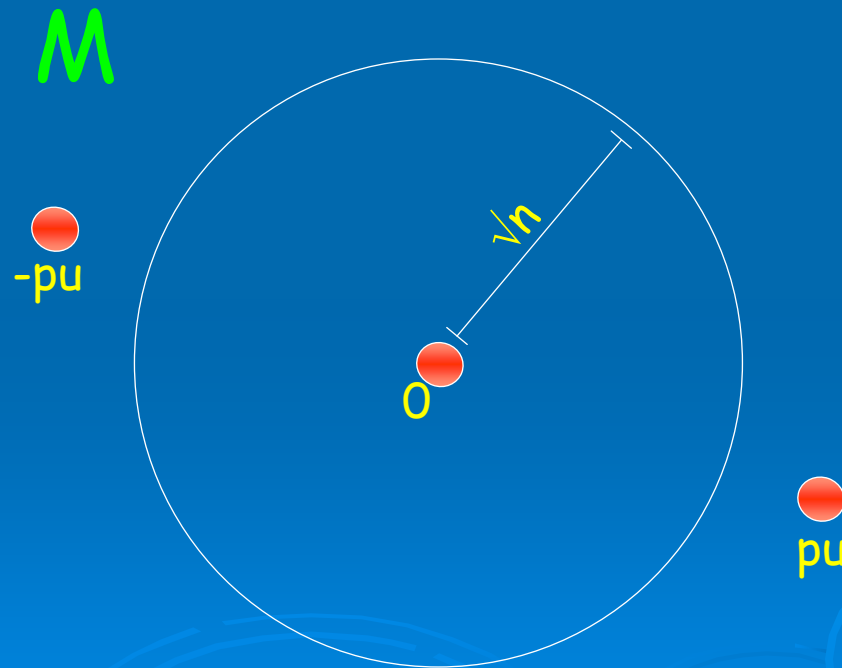
- The lattice  $M$  is spanned by  $pv_1, v_2, \dots, v_n$ :
- If  $p|a_1$ , then  $u = (a_1/p) \cdot pv_1 + a_2v_2 + \dots + a_nv_n \in M$   
:



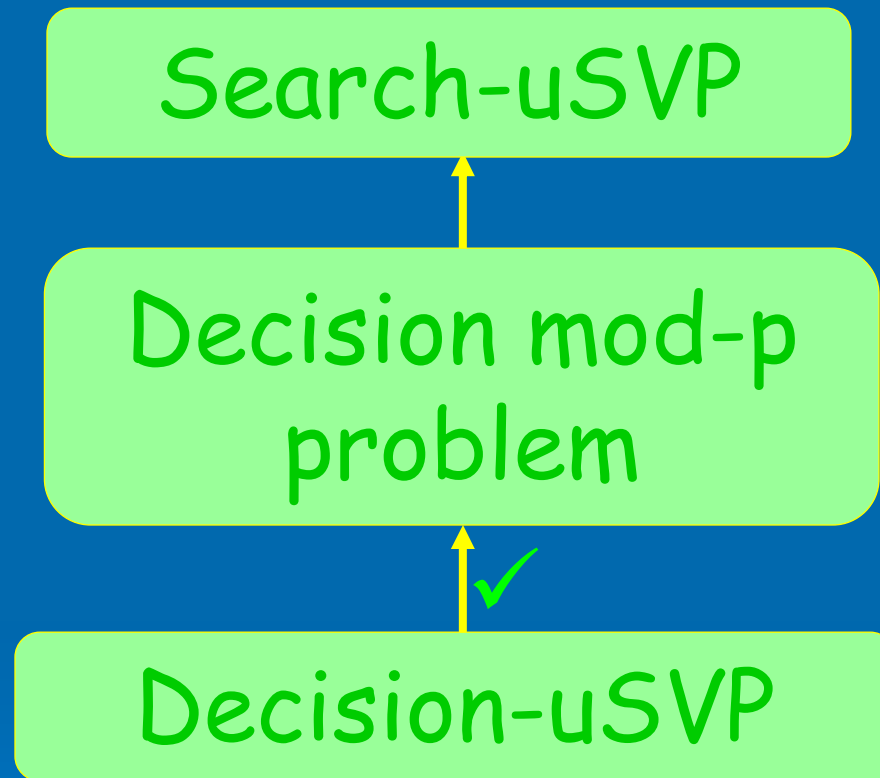
# The lattice $M$



- The lattice  $M$  is spanned by  $pv_1, v_2, \dots, v_n$ :
- If  $p \nmid a_1$ , then  $u \notin M$ :



# uSVP Decision $\rightarrow$ Search



# Reduction from: Decision mod-p



- Given a basis  $(v_1 \dots v_n)$  for  $n^{1.5}$ -unique lattice, and a prime  $p > n^{1.5}$
- Assume the shortest vector is:

$$u = a_1 v_1 + a_2 v_2 + \dots + a_n v_n$$

- Decide whether  $a_1$  is divisible by  $p$

# The Reduction



- Idea: decrease the coefficients of the shortest vector
- If we find out that  $p|a_1$  then we can replace the basis with  $p\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_n$ .
- $\mathbf{u}$  is still in the new lattice:

$$\mathbf{u} = (a_1/p) \cdot p\mathbf{v}_1 + a_2\mathbf{v}_2 + \dots + a_n\mathbf{v}_n$$

- The same can be done whenever  $p|a_i$  for some  $i$



# The Reduction

- But what if  $p \nmid a_i$  for all  $i$ ?
- Consider the basis  $v_1, v_2 - v_1, v_3, \dots, v_n$
- The shortest vector is
 
$$u = (a_1 + a_2)v_1 + a_2(v_2 - v_1) + a_3v_3 + \dots + a_nv_n$$
- The first coefficient is  $a_1 + a_2$
- Similarly, we can set it to
 
$$a_1 - bp/2ca_2, \dots, a_1 - a_2, a_1, a_1 + a_2, \dots, a_1 + bp/2ca_2$$
- One of them is divisible by  $p$ , so we choose it and continue



# The Reduction

- Repeating this process decreases the coefficients of  $u$  by a factor of  $p$  at a time
  - The basis that we started from had coefficients  $\leq 2^{2n}$
  - The coefficients are integers
- ➔ After  $\leq 2n^2$  steps, all the coefficient but one must be zero
- The last vector standing must be  $\pm u$

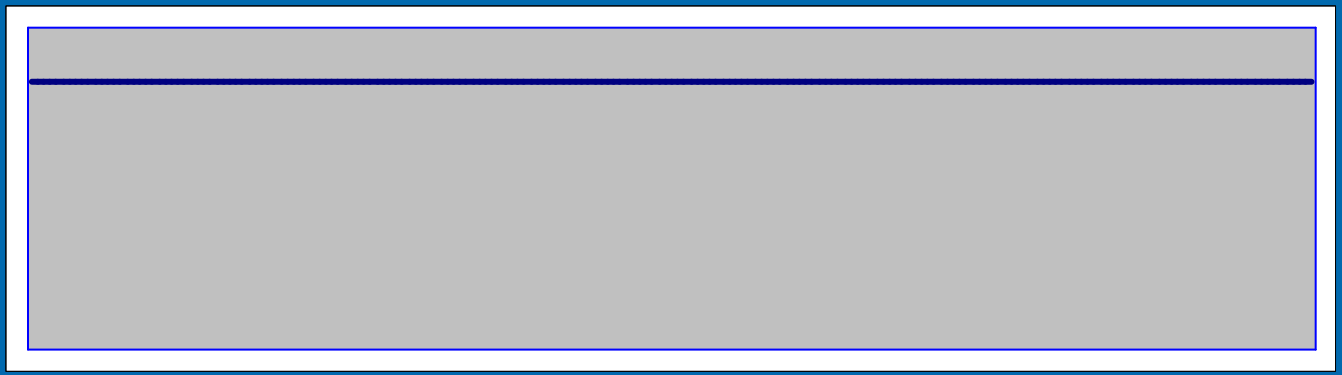


# Regev's dimension reduction

# Reducing from n to 1-dimension

- Distinguish between the 1-dimensional distributions:

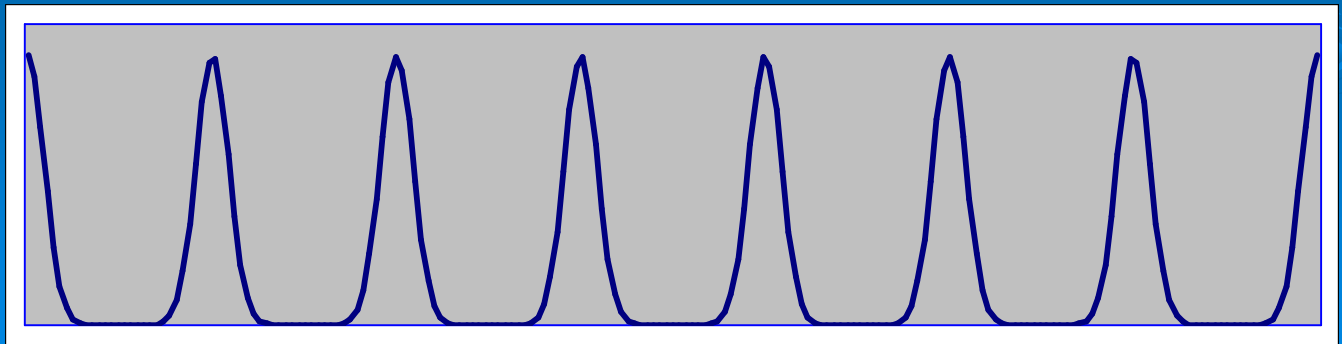
Uniform:



0

R-1

Wavy:

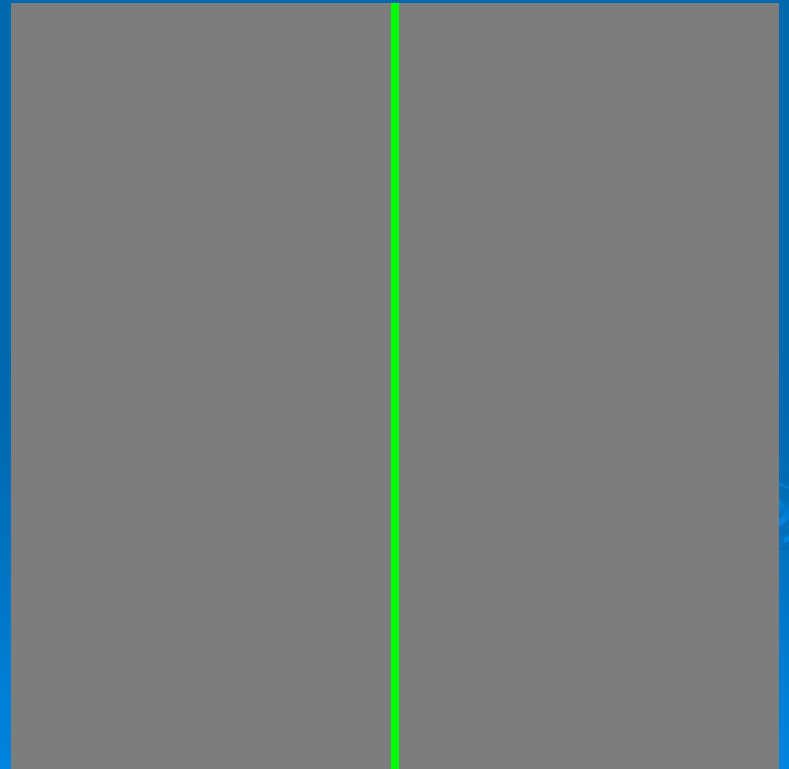
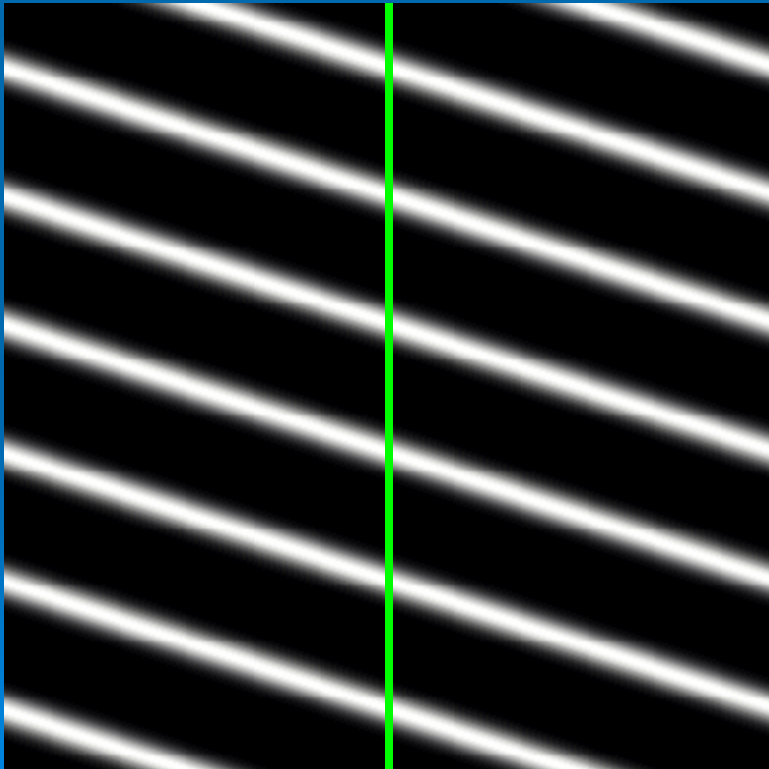


0

60  
R-1

# Reducing from $n$ to 1-dimension <sup>®</sup>

- First attempt: sample and project to a line



# Reducing from $n$ to 1-dimension

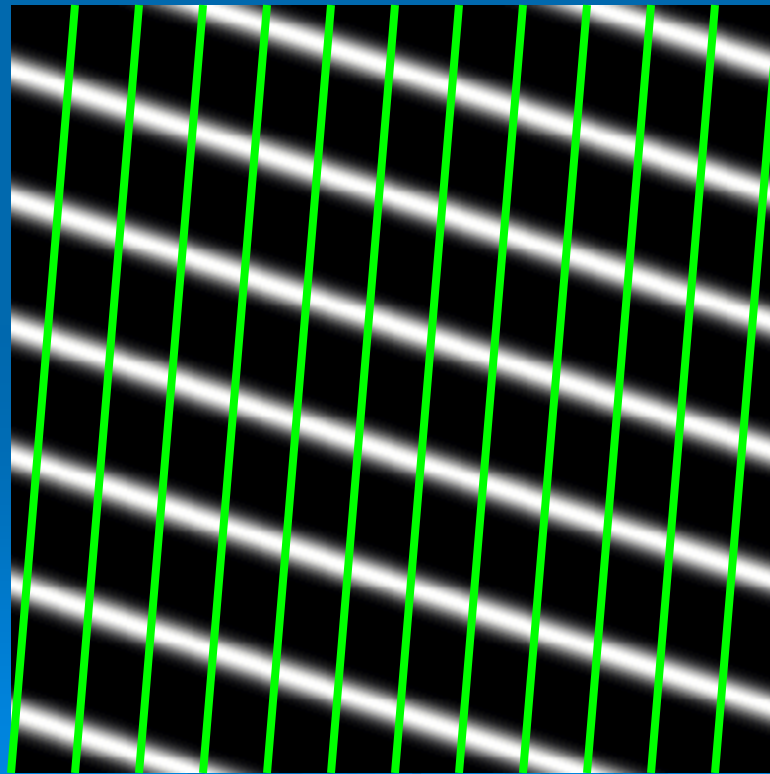
- But then we lose the wavy structure!
- We should project only from points very close to the line



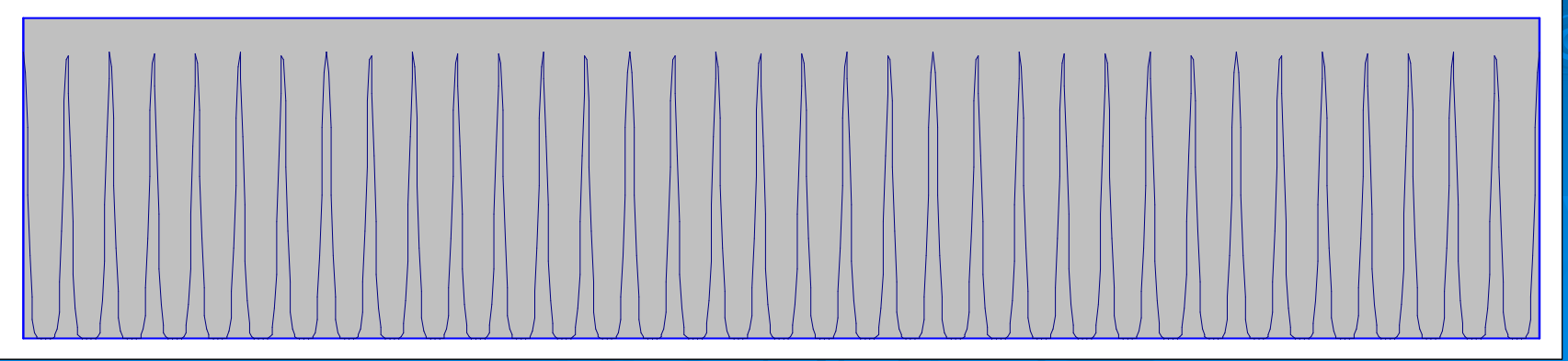
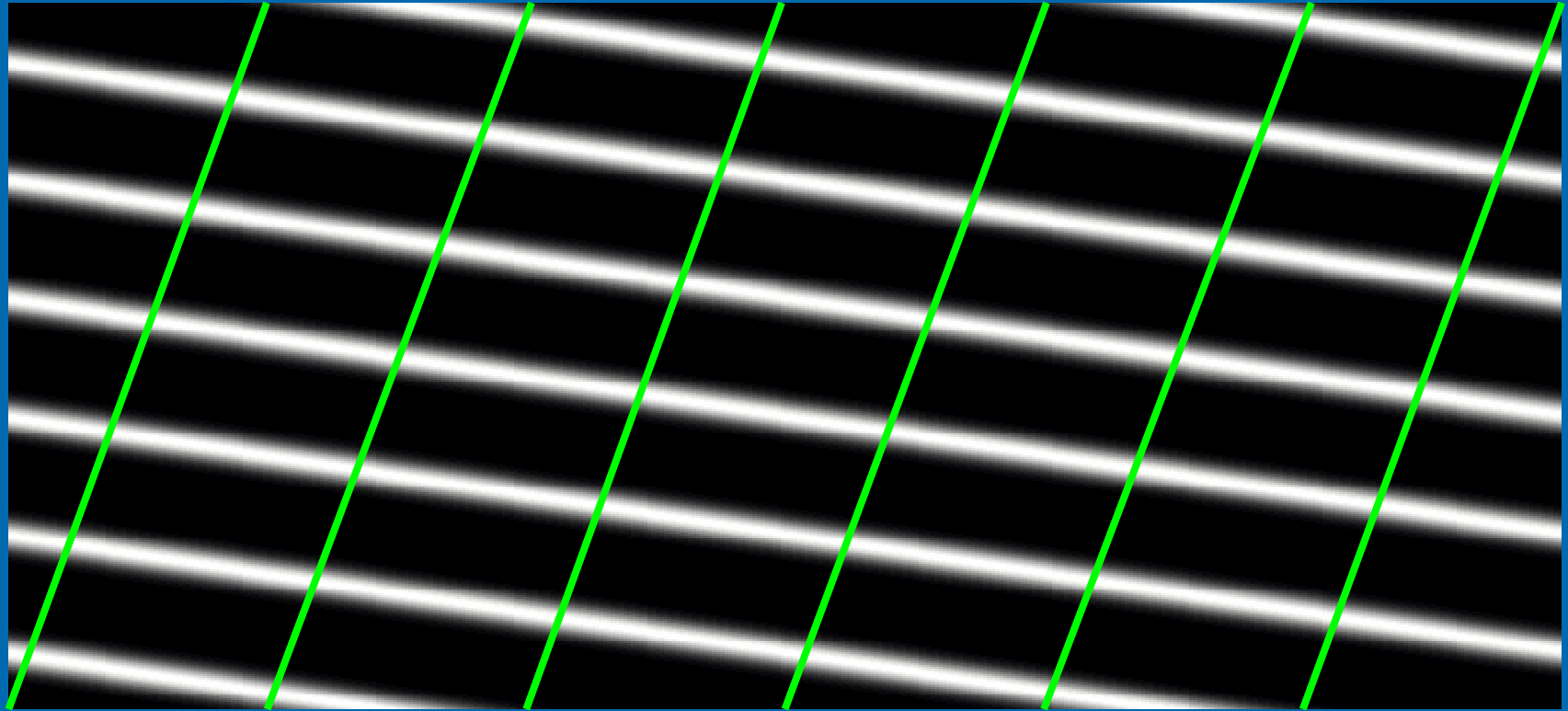
# The solution



- Use the periodicity of the distribution
- Project on a 'dense line' :



# The solution





# The solution



- We choose the line that connects the origin to  $e_1 + Ke_2 + K^2e_3 \dots + K^{n-1}e_n$  where  $K$  is large enough
- The distance between hyperplanes is  $n$
- The sides are of length  $2^n$
- Therefore, we choose  $K = 2^{O(n)}$
- Hence,  $d < O(K^n) = 2^{O(n^2)}$

# Worst-case vs. Average-case <sup>®</sup>

- So far: a problem that is hard in the worst-case: distinguish between uniform and  $d, \gamma$ -wavy distributions for **all** integers  $d < 2^{(n^2)}$
- For cryptographic applications, we would like to have a problem that is hard on the average: distinguish between uniform and  $d, \gamma$ -wavy distributions for **a non-negligible fraction** of  $d$  in  $[2^{(n^2)}, 2 \cdot 2^{(n^2)}]$

# Compressing <sup>®</sup>

- The following procedure transforms  $d, \gamma$ -wavy into  $2d, \gamma$ -wavy for all integer  $d$ :
  - Sample  $a$  from the distribution
  - Return either  $a/2$  or  $(a+R)/2$  with probability  $\frac{1}{2}$
- In general, for any real  $\alpha \geq 1$ , we can compress  $d, \gamma$ -wavy into  $\alpha d, \gamma$ -wavy
- Notice that compressing preserves the uniform distribution
- We show a reduction from worst-case to average-case

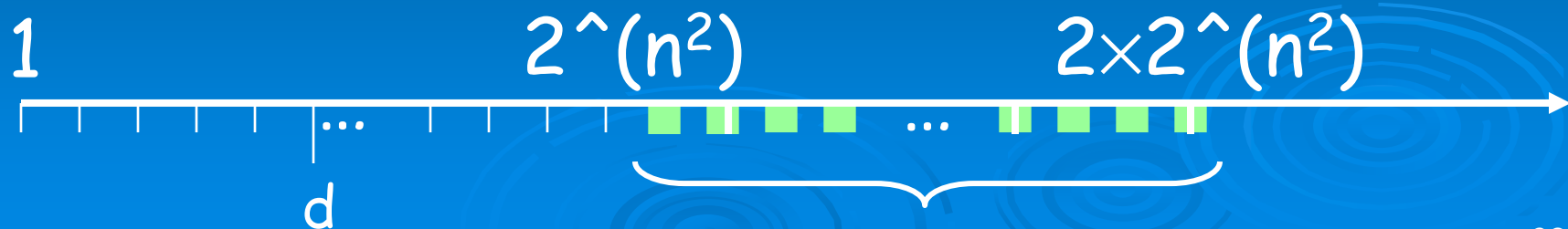
# Reduction

- Assume there exists a distinguisher between uniform and  $d, \gamma$ -wavy distribution for some non-negligible fraction of  $d$  in  $[2^{(n^2)}, 2 \cdot 2^{(n^2)}]$
- Given either a uniform or a  $d, \gamma$ -wavy distribution for some integer  $d < 2^{(n^2)}$  repeat the following:
  - Choose  $\alpha$  in  $\{1, \dots, 2 \times 2^{(n^2)}\}$  according to a certain distribution
  - Compress the distribution by  $\alpha$
  - Check the distinguisher's acceptance probability
- If for some  $\alpha$  the acceptance probability differs from that of uniform sequences, return 'wavy'; otherwise, return 'uniform'

# Reduction



- Distribution is uniform:
  - After compression it is still uniform
  - Hence, the distinguisher's acceptance probability equals that of uniform sequences for all  $\alpha$
- Distribution is  $d, \gamma$ -wavy:
  - After compression it is in the good range with some probability
  - Hence, for some  $\alpha$ , the distinguisher's acceptance probability differs from that of uniform sequences



# Diophantine Approximation

# Solving for u

(from slide 24)

- Recall: We have  $B=(b_1, \dots, b_n)$  and  $u'$ 
  - Shortest vector  $u \in L(B)$  is  $u = \sum \mu_i b_i$ ,  $|\mu_i| < 2^n$ 
    - Because the basis  $B$  is LLL reduced
  - $u'$  is very very close to  $u/|u|$ 
    - $u/|u| = (u' + e)$ ,  $|e| = 1/N$ ,  $N \gg 2^n$  (e.g.,  $N = 2^{n^2}$ )
- Express  $u' = \sum \xi_i b_i$  ( $\xi_i$ 's are reals)
- Set  $v_i = \xi_i / \xi_n$  for  $i=1, \dots, n-1$ 
  - $v_i$  very very close to  $\mu_i / \mu_n$  ( $v_i \cdot \mu_n = \mu_i + O(2^n/N)$ )

# Diophantine Approximation

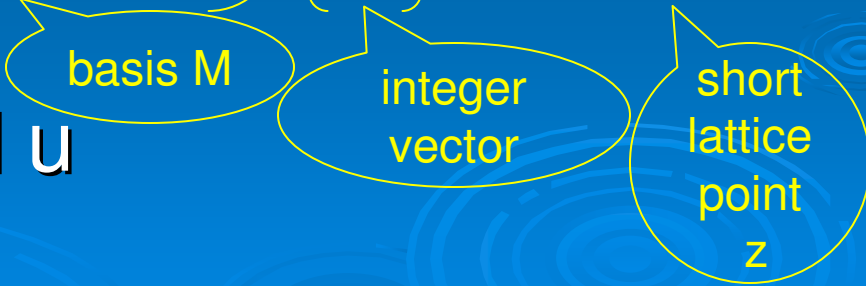
➤ Look for  $\mu_n < 2^n$  s.t. for all  $i$ ,  $v_i \cdot \mu_n$  is  $2^n/N$  away from an integer (for  $N = 2^{n^2}$ )

➤  $z$  is the unique shortest in  $L(M)$  by a factor  $\sim N/2^n$

➤ Use LLL to find it

➤ Compute the  $\mu_i$ 's and  $u$

$$\begin{pmatrix} 1 & & & & & \\ & 1 & & & & \\ & & \dots & & & \\ & & & 1 & & \\ & & & & -v_{n-1} & \\ & & & & & 1/N \end{pmatrix} \cdot \begin{pmatrix} \mu_1 \\ \mu_2 \\ \dots \\ \mu_n \end{pmatrix} = \begin{pmatrix} O(2^n/N) \\ O(2^n/N) \\ \dots \\ O(2^n/N) \\ O(2^n/N) \end{pmatrix}$$





# Why is $z$ unique-shortest?

- Assume we have another short vector  $y \in L(M)$ 
  - $\mu_n$  not much larger than  $2^n$ , also the other  $\mu_i$ 's
- Every small  $y \in L(M)$  corresponds to  $v \in L(B)$  such that  $v/|v|$  very very close to  $u'$ 
  - So also  $v/|v|$  very very close to  $u/|u|$  ( $\sim 2^n/N$ )
  - Smallish coefficient  $\rightarrow v$  not too long ( $\sim 2^{2n}$ )
  - $\rightarrow v$  very close to its projection on  $u$  ( $\sim 2^{3n}/N$ )
  - $\rightarrow \exists \chi$  s.t.  $(v - \chi u) \in L(B)$  is short
    - Of length  $\lesssim 2^{3n}/N + \lambda_1/2 < \lambda_1$
  - $\rightarrow v$  must be a multiple of  $u$

