

Autour de la réduction d'une base aléatoire

Ali Akhavi ¹ Jean-François Marckert ² Alain Rouault ³

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¹LIAFA, Université Paris VII [akhavi@liafa.jussieu.fr]

²LABRI, Université Bordeaux I [marckert@labri.fr]

³LMV, Université de Versailles-Saint-Quentin [rouault@math.uvsq.fr]

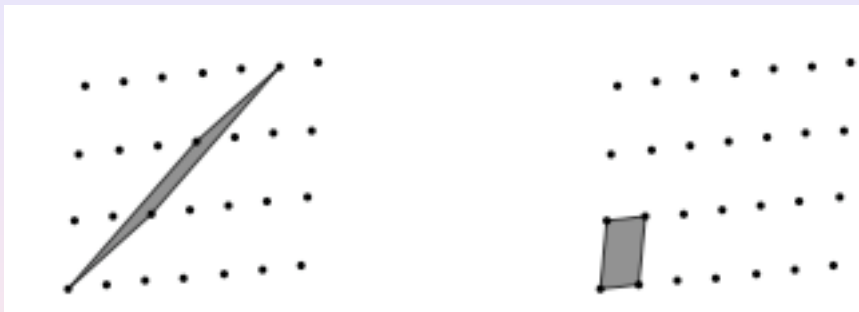
Sommaire

- 1 Le résultat : une version informelle
 - Réseau euclidien, base, le problème de la réduction
 - probabilité de réduction d'une base en fonction de la codimension
- 2 Le résultat : version plus précise
 - La LLL-réduction
 - modèle de base aléatoire
 - La probabilité asymptotique de la LLL-réduction d'une base aléatoire
- 3 Idée de la preuve

Réseau euclidien, base, Le problème de la réduction et ses motivations

- $(b) = b_1^{(n)}, b_2^{(n)}, \dots, b_p^{(n)}$ (for $p \leq n$) is a linearly independent system of p vectors of \mathbb{R}^n .
- \mathbb{R}^n with its classical Euclidean structure is the *ambient space*.
- $g := n - p$ is *the codimension* of the system (b) .
- The set $\{\sum_{i=1}^p \lambda_i b_i^{(n)} : \lambda_i \in \mathbb{Z}\}$ is an additive discrete subgroup of \mathbb{R}^n called *a lattice*.
- The system $b_1^{(n)}, b_2^{(n)}, \dots, b_p^{(n)}$ is a *basis* of the lattice.
- The lattice basis reduction problem deals with finding a basis of a given lattice, whose vectors are “short enough” and “almost orthogonal”.

Finding a nice basis of a lattice from an arbitrary basis



- On the first picture, an ugly basis, on the second, a reduced basis of the same lattice. A lattice has infinitely many bases.
- Numerous applications : cryptology, computational number theory, integer programming, computational group theory, ...
- Come to LLL+25, Caen 29,30 June, 1st and 2nd July 07 !

An informal statement of the result

- $s \in (0, 1)$ is an approximation parameter for the reducedness.
- The closer is s to 1, the better is the quality of the basis.

Theorem

Let $b_1^{(n)}, b_2^{(n)}, \dots, b_p^{(n)}$ be a random basis of codimension $g = n - p$ under some natural random model. Let $s \in (0, 1)$ be a real parameter.

(i) If $g = g(n)$ tends to infinity, then the probability that a random basis is $LLL(s)$ -reduced tends to 1.

(ii) If g is constant then the probability that a random basis is $LLL(s)$ -reduced converges to a constant in $(0, 1)$ (depending on s and g).

Gram-Schmidt orthogonalization : $(b) \longrightarrow (\widehat{b}) :$

$$\begin{cases} \widehat{b}_1 = b_1 \\ \widehat{b}_2 = b_2 \perp \langle b_1 \rangle \\ \vdots \\ \widehat{b}_p = b_p \perp \langle b_1, \dots, b_{p-1} \rangle \end{cases} \longleftrightarrow \begin{cases} \widehat{b}_j^{(n)} = b_j^{(n)} - \sum_{i=1}^{j-1} \frac{\langle b_j^{(n)}, \widehat{b}_i^{(n)} \rangle}{\|\widehat{b}_i^{(n)}\|^2} \widehat{b}_i^{(n)} \\ \text{for } j \geq 2. \end{cases}$$

$$\begin{cases} (b) = (\widehat{b}) R \\ R \text{ upper triangular :} \\ R_{j,j} = 1 \end{cases} \begin{matrix} & b_1 & \cdots & b_i & b_{i+1} & \cdots & b_p \\ \widehat{b}_1 & \left(\begin{array}{cccccc} 1 & \cdots & \cdots & \cdots & \cdots & R_{1,p} \\ \vdots & \ddots & \ddots & & & \vdots \\ \widehat{b}_i & 0 & \cdots & 1 & R_{i,i+1} & \vdots \\ \widehat{b}_{i+1} & 0 & \cdots & 0 & 1 & \ddots & \vdots \\ \vdots & \vdots & \vdots & \vdots & \ddots & \ddots & \vdots \\ \widehat{b}_p & 0 & \cdots & \cdots & \cdots & 0 & 1 \end{array} \right) \\ \vdots & & & & & & \end{matrix} \cdot$$

An LLL(s) reduced basis

$(b^{(n)}) = b_1^{(n)}, b_2^{(n)}, \dots, b_p^{(n)}$ (for $p \leq n$) is a linearly independent system of p vectors of \mathbb{R}^n , with codimension $g := n - p$.

For all $i \in \{g + 1, \dots, n - 1\}$, set : $r_i^{(n)} := \frac{\|\widehat{b}_{n-i+1}^{(n)}\|^2}{\|\widehat{b}_{n-i}^{(n)}\|^2}$

Definition

- $(b^{(n)})$ is an LLL(s)-reduced basis iff

$$\forall i \in \{1, \dots, p - 1\}, \quad \frac{\|\widehat{b}_{i+1}^{(n)}\|^2}{\|\widehat{b}_i^{(n)}\|^2} > s^2.$$

$$\Leftrightarrow \forall i \in \{g + 1, \dots, n - 1\} \quad r_i^{(n)} > s^2$$

$$\Leftrightarrow \min_{i \in \{g+1, \dots, n-1\}} r_i^{(n)} > s^2$$

Important random variables

Definition

- The reduction level of $(b^{(n)})$ is the quantity

$$\mathcal{M}_n^g := \min_{i \in \{1, \dots, p-1\}} \frac{\|\widehat{b}_{i+1}^{(n)}\|^2}{\|\widehat{b}_i^{(n)}\|^2} = \min_{i \in \{g+1, \dots, n-1\}} r_i^{(n)}$$

- The index of worst local reduction of $(b^{(n)})$ is

$$\mathcal{I}_n^g := \min \left\{ i : r_i^{(n)} = \mathcal{M}_n^g \right\} .$$

Models of random bases

- The vectors b_i 's ($1 \leq i \leq p \leq n$) are picked up randomly in \mathbb{R}^n , independently, and with the same distribution ν_n
- ν_n is invariant by rotation and satisfies $\nu_n(0) = 0$.
- A technical condition :

Assumption

There exists a deterministic sequence $(a_n)_n$ and constants $d_1, d_2, \alpha > 0, \rho_0 \in (0, 1)$ such that, for every n and $\rho \in (0, \rho_0)$

$$\nu_n \left(\left| \frac{\|b_1^{(n)}\|^2}{a_n} - 1 \right| \geq \rho \right) \leq d_1 e^{-nd_2 \rho^\alpha}.$$

In particular, $\sup \left\{ \left| \frac{\|b_i^{(n)}\|^2}{a_n} - 1 \right|, i \in \{1, \dots, n\} \right\} \xrightarrow[n]{\text{proba}} 0.$

Three examples of random bases

- ν_n is the uniform distribution on \mathbb{S}^{n-1} . In this case $\|b_1^{(n)}\|^2 = 1$, and $a_n = 1$.
- $\nu_n = \mathbb{U}_n$. In this case, $a_n = 1$ and $\mathbb{U}_n(|\|b_1^{(n)}\|^2/a_n - 1| \geq \rho) = (1 - \rho)^{n/2} \leq e^{-n\rho/2}$.
- ν_n is the n -variate standard normal (the coordinates are i.i.d. $\mathcal{N}(0, 1)$). Then $\|b_1^{(n)}\|^2/2$ is $\gamma_{n/2}$ -distributed. It can be shown that for $a_n = n$, the previous assumption holds in this case with $\alpha = 2$.

Notice that these three models are cited in the second volume of Knuth's art of programming.

The result : Some notations

- $x = (x_i)_{i \geq 1}$ a sequence of real numbers,
- $\operatorname{argmin} x = \{i : \inf_{j \geq 1} x_j = x_i\}$.
- $\|x\|_q := \left(\sum_{i \geq 1} |x_i|^q\right)^{1/q}$
- $\ell_q := \{x, \|x\|_q < +\infty\}$.

Let $(\eta_i)_{i \geq 1}$ be a sequence of independent random variables such that $\eta_i \stackrel{(d)}{=} \gamma_{i/2}$ and set

$$\mathcal{R}_k = \eta_k / \eta_{k+1}, \quad k \geq 1,$$

$$\mathcal{M}_k = \min\{\mathcal{R}_k, k \geq 1\}, \quad \text{and}$$

$$\mathcal{I}_k = \min\{\operatorname{argmin}\{\mathcal{R}_k, k \geq 1\}\},$$

The result

Proposition

For any $q > 2$, the following convergence in distribution holds in the metric space ℓ_q :

$$(r_k^n - 1)_{k \geq 1} \xrightarrow[n]{(d)} (\mathcal{R}_k - 1)_{k \geq 1}.$$

Theorem

If ν_n is spherical and satisfies Assumption 2.3 then,

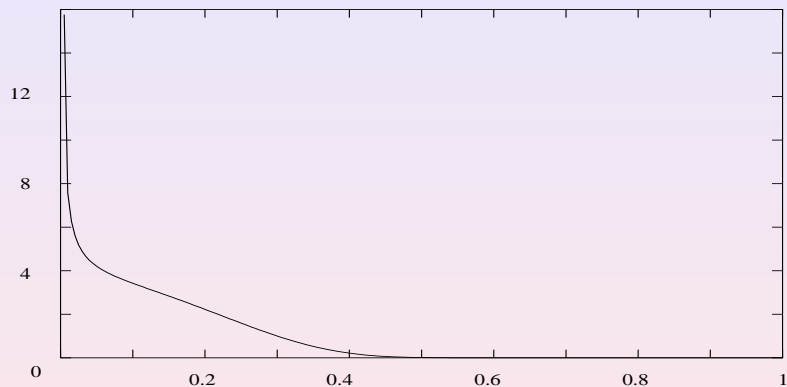
(i) Let $g : \mathbb{N} \rightarrow \mathbb{N}$ such that $g(n) \leq n$ and $g(n) \rightarrow \infty$. Then

$$\mathcal{M}_n^{g(n)} \xrightarrow[n]{\text{proba.}} 1.$$

(ii) For each $k \geq 1$, $\mathcal{M}_n^k \xrightarrow[n]{(d)} \mathcal{M}^k$.

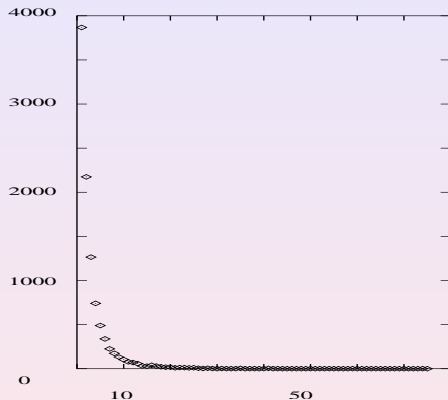
(iii) For any $k \geq 1$, $\mathcal{I}_n^k \xrightarrow[n]{(d)} \mathcal{I}^k$.

Simulation (I)



Simulation of the density of \mathcal{M}^0 with 10^8 data.

Simulation (II)



Histogram provided by 10000 simulations of \mathcal{I}_∞ .
The sequence $P(\mathcal{I}^0 = k)$ seems to be decreasing.

The Beta–Gamma algebra

- For $a > 0$, the gamma distribution of parameter a is

$$\gamma_a(dx) = \frac{e^{-x} x^{a-1}}{\Gamma(a)} \mathbb{1}_{[0, \infty)}(x) dx,$$

and its mean is a .

- For $(a, b) \in \mathbb{R}^{+*}$, the beta distribution of parameters (a, b) denoted by $\beta_{a,b}$ is

$$\beta_{a,b}(dx) = \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} x^{a-1} (1-x)^{b-1} \mathbb{1}_{(0,1)}(x) dx.$$

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$$\gamma(a) + \gamma(b) \stackrel{(d)}{=} \gamma(a+b), \quad \frac{\gamma(a)}{\gamma(b)} \stackrel{(d)}{=} \frac{\beta(a,b)}{1-\beta(a,b)},$$

-

$$\mathbb{P}(\gamma(a)/\gamma(b) \in dx) = \frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)} \frac{x^{a-1}}{(1+x)^{a+b}} \mathbb{1}_{[0, \infty)}(x) dx.$$

Idea of proof

- 1 $\|\widehat{b}_j^{(n)}\|^2 \stackrel{(d)}{=} \beta\left(\frac{n-j+1}{2}, \frac{j+1}{2}\right)$.
- 2 Convergence of the process $(r_j^{(n)})_{j \geq 1}$ with n .
 - for each j , $r_j^{(n)} \xrightarrow[n]{(d)} \gamma\left(\frac{j}{2}\right) / \gamma\left(\frac{j+1}{2}\right)$, and
 - $\gamma\left(\frac{j}{2}\right) / \gamma\left(\frac{j+1}{2}\right) \xrightarrow[j]{a.s.} 1$;
 - this suggests that the minimum \mathcal{M}_n^g is reached by the firsts $r_j^{(n)}$ and motivated the time inversions
- 3 For any $q > 2$, $(\mathcal{R}_k - 1)$ is a.s. in ℓ_q , i.e. $\sum_k |\mathcal{R}_k - 1|^q < \infty$
- 4 Definition of a process $R^{(n)}$ equidistributed to $r^{(n)}$, but defined on a unique space probability
- 5 $(R_k^{(n)} - \mathcal{R}_k)_{k \geq 1} \xrightarrow[n]{a.s.} 0$ in ℓ_q , i.e. $\sum_k |R_k^{(n)} - \mathcal{R}_k|^q \xrightarrow[n]{a.s.} 0$.
- 6 Convergence of the reduction level and related quantities

Conclusion and perspectives

- 1 A non trivial result on random lattices
 - Other properties of random lattices - ζ [Rouault]
- 2 consequences in lattice reduction :
A random basis with a high codimension is easy to reduce
 - To design efficient probabilistic reduction algorithms.
 - To analyse more precisely reduction algorithm (methods issued from dynamical systems)