

Invariant measures and random lattices

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Outline

- ▶ Invariant measures of Hurwitz and Siegel
- ▶ The fundamental domain random lattices, and lattice reduction



Hurwitz and invariant measure

Problems in **invariant theory**, in particular Cayley's finiteness problem, led Hurwitz to the notion of the invariant measure (now also referred to as the **Haar measure**) on the continuous matrix groups $SO(N)$ and $U(N)$.

The defining property is

$$\mu(G) = \mu(G_0G) = \mu(GG_0)$$

With the notation (dA) denoting the product of all independent real and imaginary parts of the differentials of the matrix A , Hurwitz observed that $(R^T dR)$ and $(U^\dagger dU)$ define the invariant measures. Thus, for example, with $R \mapsto R_0R$, $dR \mapsto R_0dR$, and so $((R_0R)^T d(R_0R)) = (R^T dR)$.

Parametrisation

Hurwitz used **Euler angles**. In the simplest cases, for $N = 2$, $R \in SO(2)$,

$$R = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

$$R^T dR = \begin{bmatrix} 0 & -d\theta \\ d\theta & 0 \end{bmatrix}, \quad (R^T dR) = d\theta$$

while for $U \in U(2)/(U(1))^2$,

$$U = \begin{bmatrix} \cos \phi & \sin \phi \\ -e^{-i\alpha} \sin \phi & e^{-i\alpha} \cos \phi \end{bmatrix},$$

and

$$U^\dagger dU = -i \begin{bmatrix} \sin^2 \phi d\alpha & i\gamma \\ -i\bar{\gamma} & \cos^2 \phi d\alpha \end{bmatrix}$$

where $\gamma = d\phi + i \sin \phi \cos \phi d\alpha$ and thus

$$(U^\dagger dU) = \sin \phi \cos \phi d\phi d\alpha.$$

Hurwitz evaluated $\int (R^T dR)$ and $\int (U^\dagger dU)$.

Parametrisation for $SO(3)$

The dimension of $SO(N)$ is $N(N - 1)/2$. Hence 3 parameters are needed for $SO(3)$. In terms of Euler angles

$$\begin{bmatrix} \cos \phi & \sin \phi & 0 \\ -\sin \phi & \cos \phi & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \theta & \sin \theta \\ 0 & -\sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} \cos \psi & \sin \psi & 0 \\ -\sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

where $0 \leq \theta \leq \pi$, $0 \leq \phi, \psi < 2\pi$.

Remark Multiplying shows the final column equals $\begin{bmatrix} \sin \theta \sin \phi \\ \sin \theta \cos \phi \\ \cos \theta \end{bmatrix}$.

This is the spherical coordinate of a unit vector in \mathbb{R}^3 with θ the polar angle, ϕ the azimuthal angle.

A calculation gives

$$(R^T dR) = \sin \theta d\theta d\phi d\psi.$$

Decomposing the invariant measure (Weyl)

Given $U \in U(N)$, from linear algebra, for $V \in U(N)/(U(1))^N$, we have the eigenvalue/ eigenvector decomposition

$$U = VDV^\dagger,$$

where $D = \text{diag}(e^{i\theta_1}, \dots, e^{i\theta_N})$.

A method based on **line elements**: $(ds)^2 = \text{Tr} dUdU^\dagger$ gives the factorisation

$$(U^\dagger dU) = \prod_{1 \leq j < k \leq N} |e^{i\theta_k} - e^{i\theta_j}|^2 d\theta_1 \cdots d\theta_N (V^\dagger dV)$$

Thus, noting $dV V^\dagger = -VdV^\dagger$ we have

$$V^\dagger dUV = V^\dagger dV D + dD - DVdV^\dagger$$

From this

$$\text{Tr} dUdU^\dagger = \sum_{j=1}^N (d\theta_j)^2 + 2 \sum_{j < k} |e^{i\theta_k} - e^{i\theta_j}|^2 (((\vec{v}_j^\dagger d\vec{v}_k)^r)^2 + ((\vec{v}_j^\dagger d\vec{v}_k)^i)^2)$$

Invariant measure for $GL_N(\mathbb{R})$ and $SL_N(\mathbb{R})$

Work of Siegel on the **geometry of numbers** lead him to consider the invariant measure on $GL_N(\mathbb{R})$,

$$d\mu(M) = \frac{(dM)}{|\det M|^N}$$

Here $(dM) = \prod_{i,j=1}^N dM_{i,j}$.

For matrices $A \in SL_N(\mathbb{R})$, Siegel defines the cone λA , $0 < \lambda < 1$, $\lambda A \in GL_N(\mathbb{R})$. From above, the latter has invariant measure equal to the Lebesgue measure (dA) . Equivalently, the invariant measure for matrices in $SL_N(\mathbb{R})$ is equal to

$$\delta(1 - \det M)(dM)$$

for $M \in GL_N(\mathbb{R})$.

Singular value decomposition

Jack and Macbeath (1959) considered the singular value decomposition

$$M = O_1 D O_2^T$$

with $O_1, O_2 \in O(N)$ and $D = \text{diag}(\sigma_1, \dots, \sigma_N)$ of the invariant measure.

Using the method of line elements they obtained the formula

$$(dM) = 2^{-N} (O_1^T dO_1) (O_2^T dO_2) \delta\left(1 - \prod_{l=1}^N \sigma_l\right) \prod_{1 \leq j < k \leq N} (\sigma_j^2 - \sigma_k^2) d\sigma_1 \cdots d\sigma_N$$

Integrating over $O_1, O_2 \in O(N)$ using Hurwitz's result leaves the task of computing the corresponding volume as

$$\int_{R > \sigma_1 > \cdots > \sigma_N > 0} \delta\left(1 - \prod_{l=1}^N \sigma_l\right) \prod_{1 \leq j < k \leq N} (\sigma_j^2 - \sigma_k^2) d\sigma_1 \cdots d\sigma_N$$

The **cutoff** R is needed since the measure is not normalisable for large singular values.

Integration method and application

Integration techniques made popular in studies of random matrix products suggest using the **Mellin transform**. Thus insert t in $\delta\left(t - \prod_{l=1}^N \sigma_l\right)$, multiply by t^s , and integrate over t to get

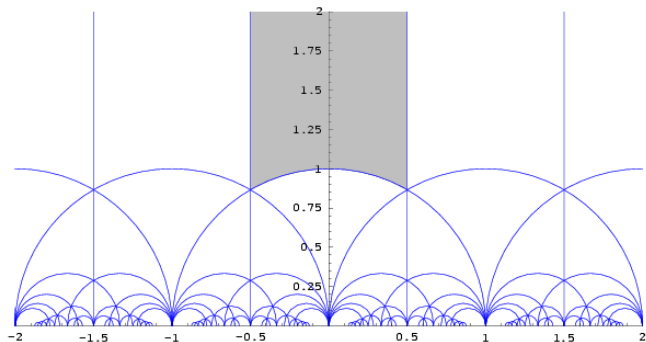
$$\frac{2^{-N}}{N!} \int_0^{R^2} dx_1 \cdots \int_0^{R^2} dx_N \prod_{l=1}^N x_l^{s/2-1} \prod_{1 \leq j < k \leq N} |x_k - x_j|$$

The **Selberg integral** gives this as a product of gamma functions. Taking the inverse Mellin transform and setting $t = 1$ implies an evaluation proportional to

$$\begin{aligned} & \frac{R^{N^2-N}}{2\pi i} \int_{c-i\infty}^{c+i\infty} R^{Ns} \prod_{j=0}^{N-1} \frac{\Gamma((s+j)/2)}{\Gamma((s+N+1+j)/2)} ds \\ & \propto \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} \left(\frac{1}{w}\right)^{[(N+1)/2]} \frac{R^{Nw}}{\prod_{r=1}^{N-1} (w^2 - (N-r)^2)^{[(r+1)/2]}} dw. \end{aligned}$$

Application: counting matrices in $SL_N(\mathbb{Z})$

Associated with the quotient $SL_N(\mathbb{R})/SL_N(\mathbb{Z})$ is a **fundamental domain** Γ and a tessellation of $SL_N(\mathbb{R})$. E.g. $N = 2$



According to Duke, Rudnick and Sarnak (1993)

$$\#\{\gamma : \gamma \in SL_N(\mathbb{Z}), \|\gamma\| \leq R\} \underset{R \rightarrow \infty}{\sim} \frac{1}{\text{vol } \Gamma} \int_{\|G\| \leq R} d\mu(G).$$

The result of Jack and Macbeath tell us the RHS for $\|G\| = \sigma_1$ (the operator norm).

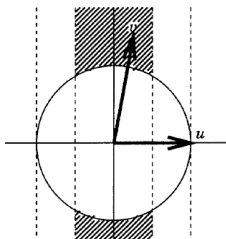
Shortest lattice vector

Basis vectors $\vec{m}_1, \dots, \vec{m}_N$. Want to choose $(n_1, \dots, n_N) \neq \vec{0}$ and $\in \mathbb{Z}^N$ such that $\left| \sum_{j=1}^N n_j \vec{m}_j \right|$ is minimum.

Question: What is the distribution of the shortest lattice vector when the basis vectors are chosen with invariant measure?

Can answer this question for $N = 2$.

For $N = 2$ it is easy to show that the shortest vector \mathbf{u} and the second shortest, linearly independent vector \mathbf{v} are characterised by the inequalities $\|\mathbf{v}\| \geq \|\mathbf{u}\|$, $2|\mathbf{u} \cdot \mathbf{v}| \leq \|\mathbf{u}\|^2$, the second being equivalent to $\|\mathbf{v} + n\mathbf{u}\| \geq \|\mathbf{v}\|$ for all $n \in \mathbb{Z}$.



QR decomposition

To align the shortest vector along the x -axis we use the QR decomposition: for $N = 2$

$$\begin{bmatrix} m_{11} & m_{12} \\ m_{21} & m_{22} \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} r_{11} & r_{12} \\ 0 & r_{22} \end{bmatrix}$$

with $r_{11} > 0$ and $r_{22} = 1/r_{11}$. Hence $\mathbf{u} = (r_{11}, 0)$ and $\mathbf{v} = (r_{12}, r_{22})$.

Invariant measure factorises according to

$$d\mu(M) = \delta(1 - \prod_{l=1}^N r_{ll}) \prod_{l=1}^N r_{ll}^{N-l} (dR)(Q^T dQ).$$

For $N = 2$, integrate over r_{22} , and $(Q^T dQ)$. Leaves $2\pi dr_{11} d_{12}$ — flat measure.

Inequalities for a reduced lattice read $r_{12}^2 + r_{22}^2 \geq r_{11}^2$, $2|r_{12}| \leq r_{11}$.

The distributions

Set $r_{11} = s$, integrate over r_{12} . Shows that the PDF of the length of the shortest lattice vector \mathbf{u} is given by

$$\frac{12}{\pi} \left(\frac{s}{2} - \chi_{s>1}(s^2 - 1/s^2)^{1/2} \right), \quad 0 < s < (4/3)^{1/4}.$$

Set $(r_{12}^2 + 1/r_{11}^2)^{1/2} = s$. Find the PDF of the second shortest basis vector \mathbf{v} is given by

$$\frac{12}{\pi s} \left((s^4 - 1)^{1/2} \chi_{1 < s < (4/3)^{1/4}} + (2s^2(s^2 - (s^4 - 1)^{1/2}) - 1)^{1/2} \chi_{(4/3)^{1/4} < s < \infty} \right).$$

Can be illustrated by the following numerical procedure:

1. Generate random matrices M from $SL_2(\mathbb{R})$ with invariant measure, constrained so that $\|M\|_{Op} \leq R$ for some (large) R . For this use the singular value decomposition and Monte Carlo to sample the singular values.
2. Apply Lagrange–Gauss lattice reduction to the columns of M , giving the reduced basis.

Lattice reduction in one-dimension

Euclid GCD

$$a = 210, \quad b = 45$$

$$r_1 = 210 - \left\lfloor \frac{210}{45} \right\rfloor 45 = 210 - 4 \times 45 = 30$$

$$r_2 = 45 - \left\lfloor \frac{45}{30} \right\rfloor 30 = 45 - 1 \times 30 = 15$$

$$r_3 = 30 - \left\lfloor \frac{30}{15} \right\rfloor 15 = 0$$

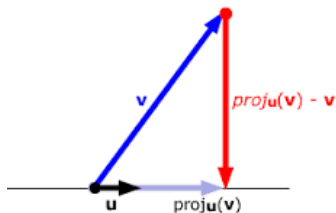
Conclusion: $\text{GCD}(210, 45) = 15$

Bézout identity

$$15 = 45 - 1 \times 30 = 45 - 1 \times (210 - 4 \times 45) = -1 \times 210 + 5 \times 45$$

Lattice reduction in two-dimensions: Lagrange-Gauss algorithm

Start with $\|\mathbf{u}\| < \|\mathbf{v}\|$. Apply orthogonal projection:



Alter $\text{proj}_{\mathbf{u}}(\mathbf{v})$ to equal $\left\lfloor \frac{\mathbf{v} \cdot \mathbf{u}}{\mathbf{u} \cdot \mathbf{u}} \right\rfloor \mathbf{u}$, and compute

$$\mathbf{q} = \mathbf{v} - \left\lfloor \frac{\mathbf{v} \cdot \mathbf{u}}{\mathbf{u} \cdot \mathbf{u}} \right\rfloor \mathbf{u}$$

If $\|\mathbf{u}\| < \|\mathbf{q}\|$ stop and output reduced basis $\{\mathbf{u}, \mathbf{q}\}$. Else, repeat with input \mathbf{u} and \mathbf{q} .

Siegel's mean value theorem

Siegel used the concept of a fundamental domain to prove a mean value theorem:

$$\left\langle \sum_{\substack{\text{lattice points} \\ \vec{y} \setminus \vec{0}}} f(y_1, \dots, y_N) \right\rangle_{\text{random lattice}} = \int_{\mathbb{R}^N} f(x_1, \dots, x_N) d\vec{x}$$

Consequence Taking $f = \chi_{R-|\vec{y}|}$ implies

$$\left\langle \# \text{ lattice points in } B_R \right\rangle = \text{vol}(B_R)$$

From this, with R_0 such that $\text{vol}(B_{R_0}) = 2$, there exists a lattice with shortest vector of length $R_0 \approx O(N^{1/2})$ for large N .

This is the Minkowski lower bound on the biggest shortest vector. Hence realised by a random lattice.

Shortest lattice vector \mathbf{b}_1 PDF and Siegel's mean value theorem

R_2 : length of 2nd linearly independent vector.

The punctured ball of radius $0 < R < R_2$ will then contain $\pm j\mathbf{b}_1$, where $1 \leq j \leq \left\lfloor \frac{R_2}{\|\mathbf{b}_1\|} \right\rfloor$.

For small s trial Cs^{d-1} for the PDF of \mathbf{b}_1 . Let $\Omega(R)$ equal the expected number of lattice vectors in the ball. Then

$$\Omega(R) = 2C \int_0^R \left\lfloor \frac{R}{s} \right\rfloor s^{d-1} ds = 2CR^d \int_0^1 \left\lfloor \frac{1}{s} \right\rfloor s^{d-1} ds = \frac{2CR^d \zeta(d)}{d}.$$

Siegel's mean value theorem gives $\Omega(R) = \text{Vol}(B_R)$. Hence

$C = \frac{d}{2\zeta(d)} \text{Vol}(B_R) \Big|_{R=1}$. E.g. $d = 3$:

