

# Lyapunov exponents and eigenvalues of products of random matrices

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# Lyapunov exponents

Consider a sequence  $\{Y_n, n \geq 1\}$  of *i.i.d*  $d \times d$  real random matrices with common distribution  $\mu$ . Let  $S_n = Y_n \cdots Y_1$  for all  $n \geq 1$ . Then the Lyapunov exponents are defined as follows.

## Definition

If  $\mathbf{E}(\log^+ \|Y_1\|) < \infty$  (we write  $a^+$  for  $\max(a, 0)$ ), the **first Lyapunov exponent** associated with  $\mu$  is the element  $\gamma$  of  $\mathbb{R} \cup \{-\infty\}$  defined by

$$\gamma_1 = \lim_{n \rightarrow \infty} \frac{1}{n} \mathbf{E}(\log \|S_n\|)$$

If  $d = 1$ , then  $\gamma_1 = \mathbf{E}(\log |Y_1|)$

## Fact

*Let  $Y_1, Y_2, \dots$  be i.i.d matrices in  $GL_d(\mathbb{R})$  and  $S_n = Y_n \cdots Y_1$ . If  $\mathbf{E}(\log^+ \|Y_1\|) < \infty$ , then with probability one*

$$\lim_{n \rightarrow \infty} \frac{1}{n} \log \|S_n\| = \gamma_1.$$

## Exterior powers

- The  $p$ -th exterior power  $\wedge^p M$  ( $1 \leq p \leq d$ ) is  $\binom{d}{p} \times \binom{d}{p}$  matrix.

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- Rows and columns are indexed by  $p$ -sized subsets of  $\{1, 2, \dots, d\}$  in dictionary order, such that  $I, J$ -th element is given by

$$\wedge^p M_{I,J} = [M]_{I,J},$$

where  $[M]_{I,J}$  denotes the  $p \times p$  minor of  $M$  that corresponds to the rows with index in  $I$  and the columns with index in  $J$ . ( $[M]_I$  denotes  $[M]_{I,I}$ ).

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- By Cauchy-Binet formula,  $[MN]_{I,J} = \sum_K [M]_{I,K} [N]_{K,J}$ . Therefore,

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- Let  $M = U\Sigma V^*$  be a singular value decomposition of  $M$  and  $\sigma_1(M) \geq \sigma_2(M) \cdots \geq \sigma_d(M)$  be singular values of  $M$ , then

$$\wedge^p M = (\wedge^p U)(\wedge^p \Sigma)(\wedge^p V^*)$$

is a singular value decomposition of  $\wedge^p M$ . Therefore

$$\|\wedge^p M\| = \sigma_1(M) \cdots \sigma_p(M).$$

## Definition

The **Lyapunov exponents**  $\gamma_1, \gamma_2 \dots \gamma_d$  associated with  $\mu$  are defined inductively by  $\gamma = \gamma_1$  and for  $p \geq 2$

$$\sum_{i=1}^p \gamma_i = \lim_{n \rightarrow \infty} \frac{1}{n} \mathbf{E}(\log \| \wedge^p S_n \|).$$

If for some  $p$   $\sum_{i=1}^{p-1} \gamma_i = -\infty$ , we put  $\gamma_p = \gamma_{p+1} = \dots \gamma_d = -\infty$ .

## Fact

Let  $Y_1, Y_2, \dots$  be i.i.d matrices in  $GL_d(\mathbb{R})$  and  $S_n = Y_n \cdots Y_1$ . Let  $\sigma_1(n) \geq \sigma_2(n) \geq \dots \geq \sigma_d(n) > 0$  are singular values of  $S_n$ . If  $\mathbf{E}(\log^+ \|Y_1\|) < \infty$ , then with probability one, for  $1 \leq p \leq d$

$$\gamma_p = \lim_{n \rightarrow \infty} \frac{1}{n} \mathbf{E}(\log \sigma_p(n)) = \lim_{n \rightarrow \infty} \frac{1}{n} \log \sigma_p(n).$$

## Definition

Let  $\mu$  (resp.  $\mu^*$ ) be the probability measure of the invertible random matrix  $Y_1$  (resp.  $Y_1^*$ ) on  $GL_d(\mathbb{R})$ . Then  $T_\mu$  is defined to be the smallest closed semigroup in  $GL_d(\mathbb{R})$  which contains the support of  $\mu$  in  $GL_d(\mathbb{R})$ . (Support of  $\mu$  in  $GL_d(\mathbb{R})$  is the set of all points in  $GL_d(\mathbb{R})$  whose every open-neighbourhood has positive measure).



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Given a subset  $T$  of  $GL_d(\mathbb{R})$ , we define the index of  $T$  as the least integer  $r$  such that there exists a sequence  $\{M_n, n \geq 0\}$  in  $T$  for which  $\|M_n\|^{-1} M_n$  converges to a rank  $r$  matrix. We say that  $T$  is **contracting** when its index is one.  $T$  is said to be  **$p$ -contracting** if  $\{\wedge^p M; M \in T\}$  is contracting.

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## Definition

$T$  is **strongly irreducible** if there does not exist a finite family of proper linear subspaces of  $\mathbb{R}^d$ ,  $V_1, V_2 \dots V_k$  such that  $M(V_1 \cup V_2 \cup \dots V_k) = (V_1 \cup V_2 \cup \dots V_k)$  for any  $M$  in  $T$ .  $T$  is  **$p$ -strongly irreducible** if  $\{\wedge^p M; M \in T\}$  is strongly irreducible.

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Define  $\ell(M) = \max(\log^+ \|M\|, \log^+ \|M^{-1}\|)$ . Observe that

$$\frac{1}{p} |\log \| \wedge^p M \| | \leq \ell(M) \text{ for all } 1 \leq p \leq d.$$

## Fact

*Let  $\mu$  be a probability measure on  $GL_d(\mathbb{R})$  such that  $\int \log^+ \|M\| d\mu(M)$  is finite. We suppose  $T_\mu$  is irreducible. Then  $\gamma_1 > \gamma_2$  if and only if  $T_\mu$  is strongly irreducible and contracting. If  $T_\mu$  is  $p$ -strongly irreducible and  $p$ -contracting for some  $p \in \{1, \dots, d-1\}$ , then  $\gamma_p > \gamma_{p+1}$ .*

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## Theorem

Let  $Y_1, Y_2, \dots$  be i.i.d random elements of  $GL_d(\mathbb{R})$  with common distribution  $\mu$  such that  $T_\mu$  is strongly irreducible and contracting. Set  $S_n = Y_n \cdots Y_1$  and let  $\lambda_1(n)$  be an eigenvalue of  $S_n$  with maximum absolute value and  $\sigma_1(n)$  be the maximum singular value of  $S_n$ , for all  $n \geq 1$ . If for some  $\tau > 0$ ,  $\mathbf{E}e^{\tau \ell(Y_1)}$  is finite, then for any  $r > 0$  with probability one

$$\lim_{n \rightarrow \infty} \frac{1}{n^r} \log \left( \frac{|\lambda_1(n)|}{\sigma_1(n)} \right) = 0.$$

## Theorem

Let  $Y_1, Y_2, \dots$  be i.i.d random elements of  $GL_d(\mathbb{R})$  with common distribution  $\mu$  such that  $T_\mu$  is  $p$ -strongly irreducible and  $p$ -contracting for all  $1 \leq p \leq d$ . Set  $S_n = Y_n \cdots Y_1$  and let  $\lambda_1(n), \lambda_2(n), \dots, \lambda_d(n)$  be the eigenvalues of  $S_n$  such that  $|\lambda_1(n)| \geq |\lambda_2(n)| \geq \dots \geq |\lambda_d(n)| > 0$  and  $\sigma_1(n), \sigma_2(n), \dots, \sigma_d(n)$  be the singular values of  $S_n$  such that  $|\sigma_1(n)| \geq |\sigma_2(n)| \geq \dots \geq |\sigma_d(n)| > 0$ , for all  $n \geq 1$ . If for some  $\tau > 0$ ,  $\mathbf{E}e^{\tau \ell(Y_1)}$  is finite, then for any  $r > 0$  with probability one

$$\lim_{n \rightarrow \infty} \frac{1}{n^r} \log \left( \frac{|\lambda_p(n)|}{\sigma_p(n)} \right) = 0,$$

for all  $1 \leq p \leq d$ .

- $\text{Tr}(S_n) = \text{Tr}(U_n \Sigma_n V_n^*) = \sum_{i=1}^d \sigma_i(n) \langle V_n e_i, U_n e_i \rangle.$

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- $|\text{Tr}(S_n)| \leq d |\lambda_1(n)| \leq d \sigma_1(n)$
- $S_{n,k} = Y_n \cdots Y_{k+1}.$  Then

$$\text{Tr}(S_n) = \text{Tr}(S_{n,k} S_{k,0})$$

$$= \sum_{i=1}^d \sum_{j=1}^d \sigma_i(n, k) \sigma_j(k, 0) \langle V_{n,k} e_i, U_{k,0} e_j \rangle \langle V_{k,0} e_j, U_{n,k} e_i \rangle.$$

## Lemma

Let  $Y_1, Y_2, \dots$  be i.i.d random elements of  $GL_d(\mathbb{R})$  with common distribution  $\mu$  such that  $T_\mu$  is strongly irreducible and contracting and for some  $\tau > 0$ ,  $\mathbf{E}e^{\tau \ell(Y_1)}$  is finite. Set  $S_n = Y_n \cdots Y_1$  and consider a Singular value decomposition  $S_n = U_n \Sigma_n V_n^*$  with  $U_n$  and  $V_n$  in  $O(d)$ , for all  $n \geq 1$ . Then for any  $r > 0$

- (i) for any sequence  $\{u_n, n \geq 1\}$  of random unit vectors in  $\mathbb{R}^d$  such that  $S_n e_1$  and  $u_n$  are independent for each  $n$ , with probability one we have that

$$\lim_{n \rightarrow \infty} \left| \frac{\langle S_n e_1, u_n \rangle}{\|S_n e_1\|} \right|^{\frac{1}{n^r}} = \lim_{n \rightarrow \infty} |\langle U_n e_1, u_n \rangle|^{\frac{1}{n^r}} = 1$$

- (ii) for any sequence  $\{v_n, n \geq 1\}$  of random unit vectors in  $\mathbb{R}^d$  such that  $S_n^* e_1$  and  $v_n$  are independent for each  $n$ , with probability one we have that

$$\lim_{n \rightarrow \infty} \left| \frac{\langle S_n^* e_1, v_n \rangle}{\|S_n^* e_1\|} \right|^{\frac{1}{n^r}} = \lim_{n \rightarrow \infty} |\langle V_n e_1, v_n \rangle|^{\frac{1}{n^r}} = 1.$$

## Corollary

*Let  $\xi$  be a real valued continuous random variable whose support contains an open set and there exists  $\tau > 0$  such that*

$$\mathbf{E}|\xi|^\tau < \infty \text{ and } \sup_{a \in \mathbb{R}} \mathbf{E}|\xi - a|^{-\tau} < \infty.$$

*Let  $X_1, X_2, \dots$  be a sequence of  $d \times d$  i.i.d random matrices whose elements are i.i.d random variables distributed like  $\xi$ . Then for any  $r > 0$  with probability one for all  $1 \leq p \leq d$ ,*

$$\lim_{n \rightarrow \infty} \frac{1}{n^r} \log \left( \frac{|\lambda_p(n)|}{\sigma_p(n)} \right) = 0.$$

## Remark

If the random variable  $\xi$  has bounded density, then  $\sup_{a \in \mathbb{R}} \mathbf{E} |\xi - a|^{-\tau} < \infty$  for any  $0 < \tau < 1$ . Let us say  $f$  is the probability density of  $\xi$ , bounded by  $K$ , then

$$\begin{aligned} \mathbf{E} |\xi - a|^{-\tau} &= \sup_{a \in \mathbb{R}} \int |x - a|^{-\tau} f(x) dx \\ &= \int_{|x-a| < \epsilon} |x - a|^{-\tau} f(x) dx + \int_{|x-a| \geq \epsilon} |x - a|^{-\tau} f(x) dx \\ &\leq K \int_{|x-a| < \epsilon} |x - a|^{-\tau} dx + \epsilon^{-\tau} \\ &= K \frac{2\epsilon^{1-\tau}}{1-\tau} + \epsilon^{-\tau} < \infty. \end{aligned}$$

Since  $\epsilon, K, \tau$  do not depend on  $a$ , we get that

$$\sup_{a \in \mathbb{R}} \mathbf{E} |\xi - a|^{-\tau} < \infty.$$

## Fact

$T_\mu$  is  $p$ -strongly irreducible and  $p$ -contracting, for any  $1 \leq p \leq d$ , if there exists a matrix  $M$  such that

- $\|M\|^{-1}M$  is not orthogonal,
- for any  $K \in O(d)$ ,  $KMK^{-1}$  is in  $T_\mu$ .

Thank you!!