

## A Characterization of Dobrushin's Coefficient of Ergodicity

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Es wird bewiesen, daß ein zu einer Vektornorm des  $\mathbb{R}^n$  gehörender Ergodizitätskoeffizient  $\tau$  genau dann die Ungleichung  $\tau(P) \leq 1$  für alle stochastischen Matrizen  $P$  der Ordnung  $n$  erfüllt, wenn er der Dobrushinsche Ergodizitätskoeffizient ist.

Доказывается, что для коэффициента эргодичности  $\tau$  относительно некоторой нормы в  $\mathbb{R}^n$  выполнено неравенство  $\tau(P) \leq 1$  для всех стохастических матриц  $P$  порядка  $n$  тогда и только тогда когда он является коэффициентом эргодичности Добружина.

It is proved that the ergodicity coefficient  $\tau$  corresponding to any vector norm on  $\mathbb{R}^n$  fulfils the inequality  $\tau(P) \leq 1$  for all  $n \times n$  stochastic matrices  $P$  iff it is the Dobrushin ergodicity coefficient.

Let  $\|\cdot\|$  be an arbitrary norm on  $\mathbb{R}^n$  and  $S_n$  ( $n \geq 2$ ) the set of all  $n \times n$  stochastic matrices. In [4] E. SENETA has introduced a general concept of coefficients of ergodicity  $\tau$  for  $P \in S_n$  with respect to  $\|\cdot\|$ :

$$\tau(P) = \sup \{ \|xP\| : x \in H, \|x\| = 1 \} \quad (P \in S_n), \quad (1)$$

where  $H = \{x = (x_1, \dots, x_n) \in \mathbb{R}^n : x_1 + \dots + x_n = 0\}$ . For the  $l_1$ -norm  $\|\cdot\|_1$ ,  $\|x\|_1 = |x_1| + \dots + |x_n|$  ( $x \in \mathbb{R}^n$ ), the ergodicity coefficient denoted by  $\tau_1$  is the well-known Dobrushin coefficient

$$\tau_1(P) = \frac{1}{2} \max_{ij} \sum_{k=1}^n |p_{ik} - p_{jk}| \quad (P \in S_n) \quad (2)$$

with  $P = (p_{ij})_{i,j=1}^n$  (see the remark following Lemma 2). The coefficients  $\tau(P)$  are bounds on all non-unit eigenvalues of  $P$ . For  $\tau_1$  the inequality  $\tau_1(P) \leq 1$  ( $P \in S_n$ ) holds. In our note we show that  $\tau_1$  is the only ergodicity coefficient  $\tau$  satisfying this inequality. The proof points out the role of certain extremal points.

Denote  $K, S, K_1$  and  $S_1$  the set of all  $x \in \mathbb{R}^n$  with  $\|x\| \leq 1$ ,  $\|x\| = 1$ ,  $\|x\|_1 \leq 1$  and  $\|x\|_1 = 1$ , respectively. For  $i, j \in \{1, \dots, n\}$ ,  $i \neq j$ , let  $e_{ij} = (x_1, \dots, x_n)$  with  $x_i = -x_j = 1/2$  and  $x_l = 0$  for  $l \neq i, j$ , and denote  $E_1 = \{e_{ij}\}_{ij}$ . Let  $x^+ = \max(x, 0)$  and  $x^- = \max(-x, 0)$  for  $x \in \mathbb{R}$ . Finally, for a linear subspace  $L \subset \mathbb{R}^n$ ,  $L \neq \{0\}$ , and norms  $|\cdot|_i$  on  $L$  denote  $B_i$  the corresponding unit balls and  $\text{Ex } B_i$  the set of their extremal points ( $i = 1, 2$ ).

Lemma 1: For  $\lambda > 0$  the following assertions are equivalent:

- (i)  $|x|_1 = \lambda |x|_2$  ( $x \in L$ );    (ii)  $\lambda = \sup_{B_1} |x|_1$  and  $|e|_2 = 1/\lambda$  ( $e \in \text{Ex } B_1$ ).

Proof: The implication (i)  $\Rightarrow$  (ii) is obvious. (ii)  $\Rightarrow$  (i): The relation  $\lambda = \sup_{B_1} |x|_1$  yields  $|x|_1 \leq \lambda |x|_2$  ( $x \in L$ ). On the other hand, if  $x \in L$  by the Krein-Milman Theorem, there are  $e_i \in \text{Ex } B_1$  and  $\lambda_i \in \mathbb{R}$  ( $i = 1, \dots, k$ ) with  $\lambda_i \geq 0$ ,  $\lambda_1 + \dots + \lambda_k = 1$  such that  $x = |x|_1 \sum_i \lambda_i e_i$ . Thus, by  $|e_i|_2 = 1/\lambda$  the inequality  $|x|_2 \leq (1/\lambda) |x|_1$  follows ■

**Lemna 2:** *The equality  $\text{Ex}(K_1 \cap H) = E_1$  is true.*

**Proof:** Clearly,  $\text{Ex}(K_1 \cap H) \subset S_1 \cap H$ . Since  $x = \left(\sum_i x_i^+\right)^{-2} \sum_{ij} x_i^+ x_j^- e_{ij}$  ( $x \in S_1 \cap H$ ), each  $x \in S_1 \cap H$  is a convex combination of elements of  $E_1$ . Therefore, we have  $\text{Ex}(K_1 \cap H) \subset E_1$ . On the other hand, let  $e_{kl} = \lambda x + (1 - \lambda) y$  with  $x, y \in K_1 \cap H$  and  $0 < \lambda < 1$ . Since  $\lambda x_k + (1 - \lambda) y_k = 1/2$ ,  $\lambda x_l + (1 - \lambda) y_l = -1/2$  and  $\sum_i x_i^+ = \sum_i x_i^- = 1/2 \|x\|_1 \leq 1/2$ ,  $\sum_i y_i^+ = \sum_i y_i^- = 1/2 \|y\|_1 \leq 1/2$  we have  $x_k = y_k = 1/2$ ,  $x_l = y_l = -1/2$ , and therefore  $x = y = e_{kl}$ . Thus,  $E_1 \subset \text{Ex}(K_1 \cap H)$  ■

**Remark:** Using Lemma 2 and (1) one can easily prove (2). Indeed, for each  $P \in S_n$  the norm  $\| \cdot P \|_1$  is a convex functional on  $K_1 \cap H$  assuming its maximum at a point of  $\text{Ex}(K_1 \cap H)$ .

Therefore, one obtains  $\tau_1(P) = \sup_{K_1 \cap H} \|xP\|_1 = \max_{E_1} \|eP\|_1 = \frac{1}{2} \max_{ij} \sum_{k=1}^n |p_{ik} - p_{jk}|$ .

**Theorem:** *Let  $\tau$  be an ergodicity coefficient with respect to the norm  $\| \cdot \|$  satisfying  $\tau(P) \leq 1$  for all  $P \in S_n$ . Then*

(i)  $\tau(P) = \tau_1(P)$  for all  $P \in S_n$ ; (ii)  $\lambda \|x\| = \|x\|_1$  ( $x \in H$ ) for some  $\lambda > 0$ .

**Proof:** Since  $\| \cdot \|_1$  is a continuous functional on the compact set  $S \cap H$ , there is a  $y \in S \cap H$  with  $\|y\|_1 = \max_{S \cap H} \|x\|_1 = \lambda$ . For  $j, k \in \{1, \dots, n\}$ ,  $j \neq k$ , let  $P_{j,k} = (p_{mi}) \in S_n$  be defined by  $p_{mj} = 1$  for  $y_m^+ > 0$  and  $p_{mk} = 1$  for  $y_m^+ = 0$ . Then  $yP_{j,k} = 2 \sum_i y_i^+ e_{jk} = \|y\|_1 e_{jk}$ . Therefore,  $\tau(P) \leq 1$  ( $P \in S_n$ ) yields  $\|e_{jk}\| \leq 1/\|y\|_1 = 1/\lambda$ . By definition of  $\lambda$ ,  $1 = \|e_{jk}\|_1 \leq \lambda \|e_{jk}\|$ , so that  $\|e_{jk}\| = 1/\lambda$ . Thus, Lemmata 2 and 1 imply (ii), and (i) directly follows ■

**Remark:** After we had finished the first form of this paper we has been informed by A. Lešanovský that he has obtained independently the result of the theorem [2]. However his proof does not use the aspect of extremal points.

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