

# THE GEOMETRY OF FINITE MARKOV CHAINS

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The purpose of this paper is to present a geometric theorem which provides a proof of a fundamental theorem of finite Markov chains.

The theorem, stated in matrix theoretic terms, concerns the asymptotic behaviour of the powers of an  $n$  by  $n$  stochastic matrix, that is, a matrix of non-negative entries each of whose row sums is 1. The matrix might arise from a repeated physical process which goes from one of  $n$  possible states to another at each iteration and whose probability of going to a state depends only on the state it is in at present and not on its more distant history. The entry  $a_{ij}$  of the matrix  $A$  (called the one step transition matrix) is the probability that the process goes from state  $i$  to state  $j$  in one step. The  $ij$ -th entry in  $A^m$ , which is denoted by  $a_{ij}^{(m)}$ , is the probability of going from  $i$  to  $j$  in  $m$  steps. For example the process might consist of shuffling a deck of  $n$  cards by means of a machine which puts the  $i$ -th card from the top into the  $j$ -th from the top with probability  $a_{ij}$ . Then  $a_{ij}^{(m)}$  is the probability of finding the  $i$ -th card in the  $j$ -th position at the  $m$ -th shuffle.

For each  $m > 0$ , we say that  $i$  leads to  $j$  in  $m$  steps iff  $a_{ij}^{(m)} > 0$ . We write  $i \rightsquigarrow j$  iff  $i$  leads to  $j$  in  $m$  steps for some  $m$ .

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It is easy to see that  $\sim$  is an equivalence relation on the set of those states on which  $\sim$  is symmetric, i. e. on

$$E = \bigcap_{j=1}^n \{i : i \sim j \text{ implies } j \sim i\}. \quad E \text{ is called the set of}$$

ergodic states.  $E$  is partitioned by  $\sim$  into  $\nu$  equivalence classes  $E_1, E_2, \dots, E_\nu$  called ergodic classes. The states not in  $E$  are called transient.

Theorems I and II below are respectively probabilistic and matrix theoretic statements of the fundamental theorem of finite Markov chains. We shall provide a geometric proof at the conclusion of the paper. For probabilistic proofs we refer the reader to [1], [2], [3], and [4]. An algebraic proof can be found in [5].

**THEOREM I.** If  $A = (a_{ij})$  is the one step transition matrix of a Markov chain with  $n$  states then:

$$(Ia) \lim_{m \rightarrow \infty} a_{ij}^{(m)} = 0 \quad \text{whenever } j \text{ is transient.}$$

There is a partition of each  $\sim$  equivalence class  $E_r$  into  $c_r$  non-empty subsets (called cyclically moving classes)  $E_{r0}, E_{r1}, \dots, E_{rc_r-1}$  with the following properties:

(Ib) If  $i \in E_{rs}$  and  $i$  leads to  $j$  in one step then  $j \in E_{r, s+1}$  (the second subscript is read modulo  $c_r$ ).

(Ic) To each  $E_{rs}$  corresponds an  $n$ -tuple  $k_{rs}$  of non-negative numbers whose sum is 1 for which the  $j$ -th component,  $k_j^{(r, s)}$ , is zero iff  $j \notin E_{rs}$  and such that:

$$\lim_{m \rightarrow \infty} a_{ij}^{(mc_r + t)} = k_j^{(r, s')} \quad \text{for all } j$$

and all  $t = 0, 1, \dots, c_r - 1$  whenever  $i \in E_{rs}$  and  $s' \equiv s + t \pmod{c_r}$ .

**THEOREM II.** If  $A = (a_{ij})$  is an  $n$  by  $n$  stochastic matrix then there is a permutation matrix  $P$  (i.e. a matrix of zeroes and ones which has only one non-zero entry in each row and each column) such that:

$$PAP^{-1} = \begin{vmatrix} A_{00} & A_{01} & A_{02} & \dots & A_{0r} & \dots & A_{0, \nu-1} & A_0 \\ 0 & A_1 & 0 & \dots & 0 & \dots & 0 & 0 \\ \cdot & & & & & & & \\ \cdot & & & & & & & \\ \cdot & & & & & & & \\ 0 & 0 & 0 & \dots & A_r & \dots & 0 & 0 \\ 0 & 0 & 0 & \dots & 0 & \dots & A_{\nu-1} & 0 \\ 0 & 0 & 0 & \dots & 0 & \dots & 0 & A_\nu \end{vmatrix} \quad (\text{if } E \neq \emptyset),$$

$$PAP^{-1} = \begin{vmatrix} A_1 & 0 & \cdot & \cdot & \cdot & 0 \\ 0 & A_2 & \cdot & \cdot & \cdot & 0 \\ 0 & \cdot & \cdot & \cdot & A_r & \cdot & \cdot & 0 \\ 0 & \cdot & \cdot & \cdot & \cdot & 0 & A_\nu \end{vmatrix} \quad (\text{if } E = \emptyset),$$

$$A_r = \begin{pmatrix} 0 & A_{r1} & 0 & 0 & \dots & 0 & 0 \\ 0 & 0 & A_{r2} & 0 & \dots & 0 & 0 \\ \cdot & & & & & & \\ \cdot & & & & & & \\ \cdot & & & & & & \\ 0 & 0 & 0 & 0 & \dots & A_{rs} & \dots & 0 \\ \cdot & & & & & & & \\ \cdot & & & & & & & \\ \cdot & & & & & & & \\ 0 & 0 & 0 & 0 & \dots & 0 & A_{rc_r-1} \\ A_{r0} & 0 & 0 & 0 & \dots & 0 & 0 \end{pmatrix},$$

where the  $A_r$  and  $A_{00}$  are square matrices and:

$$(IIa) \lim_{m \rightarrow \infty} A_{00}^m = 0,$$

(IIb) the entries of  $A_r$  which are in no  $A_{rs}$  are zero,

(IIc) for each  $r = 1, 2, \dots, \nu$  there are stochastic matrices  $\Pi_{r0}, \Pi_{r1}, \dots, \Pi_{rs}, \dots, \Pi_{r,c_r-1}$  such that for each  $t = 0, 1, \dots, c_r - 1$ :

$$\lim_{m \rightarrow \infty} A_r^{mc_r+t} = \begin{array}{ccccccc} 0 & \dots & 0 & \Pi_{rt} & 0 & \dots & 0 \\ 0 & \dots & 0 & & \Pi_{r,t+1} & \dots & 0 \\ & & & & & \cdot & \\ & & & & & \cdot & \\ & & & & & \cdot & \\ 0 & \dots & & & & 0 & \Pi_{r,c_r-1} \\ \Pi_{r0} & 0 & & \dots & & & 0 \\ 0 & \Pi_{r1} & 0 & & \dots & & 0 \\ & & \cdot & & & & \cdot \\ & & \cdot & & & & \cdot \\ & & \cdot & & & & \cdot \\ 0 & \dots & 0 & \Pi_{r,t-1} & 0 & \dots & 0 \end{array}$$

The entries in this matrix are zero if and only if they are in no  $\Pi_{rs}$ , there are as many rows in  $\Pi_{rs}$  as there are rows in  $A_{r,s-t+1}$  (second subscript modulo  $c_r$ ), and all of the rows of  $\Pi_{rs}$  are the same vector  $\pi^{(r,s)}$ .

The method we shall use to prove the fundamental theorem and related theorems is briefly this: we identify the  $n$  by  $n$  stochastic matrix  $A$  with a linear operator  $f$  on the simplex,  $S$ , spanned by the basis vectors of Euclidean  $n$ -space. The intersection,  $K$ , of all the images  $f^m(S)$  is a simplex whose vertices are permuted by  $f$ . The position of  $K$  in  $S$  and this permutation determine the behaviour of the  $a_{ij}^{(m)}$  for large  $m$  and also locate the vertices of the simplex of its stochastic eigenvectors.

Before going further we shall state a few definitions and preliminary remarks for the reader's convenience.

A convex polytope  $P$  is the convex hull of finitely many points  $t_1, t_2, \dots, t_m$  in some Euclidean  $n$ -space. The point  $t_1$  is a vertex of  $P$  iff the convex hull of the others doesn't contain it. The convex hull of any subset of the vertices of  $P$  is called a sub-polytope of  $P$ . A linear function  $f$  mapping  $P$  into  $P$  is called a linear operator on  $P$ . A convex polytope  $P$  is a simplex iff none of its vertices is in the flat determined by the remaining vertices. If  $S$  is a simplex, each subpolytope is called a subsimplex. The subsimplices of  $S$  are themselves simplices. If  $A$  is a subset of the convex polytope  $P$ , a carrier of  $A$  in  $P$  is a subpolytope with fewest vertices, containing  $A$ . If  $P$  is a simplex then each non empty subset has a unique carrier in  $P$ .

Three direct consequences of these definitions which we shall refer to in the sequel are:

- (1) subsimplices without vertices in common are disjoint;
- (2) if the carriers of  $m$  points in a simplex  $S$  are disjoint then the convex hull of these points is a simplex;
- (3) if  $f$  is a linear operator on a simplex  $S$ , then the image of the carrier of a subset  $X$  of  $S$  is contained in the carrier of  $f(X)$ .

The method we use is based on a lemma which we couldn't find in the literature:

LEMMA 1. The intersection  $K$ , of a nested sequence of convex polytopes  $\{P_\alpha\}$  each of which has  $n$  vertices is a convex polytope.

Proof. It is possible to choose a subsequence,  $\{P_{\alpha_\beta}\}$  and a vertex  $v_\beta$  of  $Q_\beta = P_{\alpha_\beta}$  such that  $\{v_\beta\}$  converges, to  $k_1$  say. Next choose a subsequence  $\{Q_{\beta_\gamma}\}$  of  $\{Q_\beta\}$  and a sequence of vertices  $w_\gamma$  of  $R_\gamma = Q_{\beta_\gamma}$  with  $w_\gamma \neq v_{\beta_\gamma}$  such that

$w_y$  converges, to  $k_2$  say. And so on, getting  $k_1, k_2, \dots, k_n$ . This process must halt in  $n$  steps because the  $P_\alpha$  have only  $n$  vertices apiece. Let  $T$  be the convex hull of the  $k_i$ .

Clearly  $T \subseteq K$ . Suppose  $x \in K \sim T$ . Let  $h$  be a hyperplane separating  $x$  from  $T$ . Let  $\epsilon$  be the distance from  $h$  to  $T$ . For each  $i$  there are infinitely many  $\alpha$  for which a vertex of  $P_\alpha$  is in the sphere of radius  $\epsilon/2$  about  $k_i$ . There is therefore a member of  $\{P_\alpha\}$  on the side of  $h$  opposite  $x$  hence  $\bigcap P_\alpha$  and  $\{x\}$  are disjoint. Thus  $K \sim T = \emptyset$  and hence  $K = T$ .

LEMMA 2. If  $f$  is a continuous function mapping the compact set  $P$  into  $P$  and  $K = \bigcap_{m \geq 1} f^m(P)$  then  $f(K) = K$ .

Proof. It is sufficient to show that  $K \subseteq f(K)$ . If  $x \in K$  then  $x \in f^m(P)$  for all  $m > 0$  and hence  $x = f(x_{m-1})$  for some  $x_{m-1} \in f^{m-1}(P)$ . The  $x_m$  have a convergent subsequence  $\{x_{m_i}\}$  converging to a point  $y$  of  $P$ . If we can show that  $y \in K$  then we are through because  $x = \lim_{i \rightarrow \infty} f(x_{m_i}) = f(y)$ .

If  $y$  were not in  $K$  then, for some  $N$ ,  $y$  would not be in  $f^N(P)$ . The complement of  $f^N(P)$  contains no  $x_{m_i}$  for  $m_i > N$ . But the complement of  $f^N(P)$  is an open neighborhood of  $y$ . Therefore  $y \in K$ .

THEOREM 1. If  $f$  is a linear operator on a simplex  $S$  then

- (i) the intersection,  $K$ , of the iterates  $f^m(S)$  is a simplex, and
- (ii) the vertices of  $K$  are permuted by  $f$  and hence fall into  $\nu$  disjoint classes on each of which  $f$  is a cyclic permutation.

tion so that for  $r = 1, 2, \dots, \nu$  and  $s = 0, 1, \dots, c_r - 1$ , (where  $c_r$  is the number of elements in the  $r$ -th class) we have

(iii)  $f(k_{rs}) = k_{r, s+1}$  (the second subscript is read modulo  $c_r$ ).

If  $C_{rs}$  is the carrier of  $k_{rs}$  in  $S$  then

(iv) the  $C_{rs}$  are disjoint,

(v)  $f^t(C_{rs}) \subseteq C_{r, s+1}$ , and

(vi)  $\bigcap_{m \equiv t} f^m(C_{rs}) = k_{rs'}$  when  $s' \equiv t + s \pmod{c_r}$ .

If  $K_r$  is the subsimplex of  $K$  whose vertices are  $k_{r0}, k_{r1}, \dots, k_{r, c_r - 1}$  then

(vii) the  $K_r$  are disjoint and

(viii) the set of all  $f$ -fixed points in  $S$  is a simplex whose vertices are the barycenters of the  $K_r$ .

Proof.

(a) Lemmas 1 and 2 establish that  $K$  is a convex polytope and that  $f(K) = K$ .

(b)  $f$  permutes the vertices of  $K$ .

Let  $k$  be a vertex of  $K$ ,  $X_k = [f^{-1}(k)] \cap K$  and  $C_K(X_k)$  denote a carrier of  $X_k$  in  $K$ . Then  $f(C_K(X_k)) = \{k\}$  by remark (3) and hence  $C_K(X_k) = X_k$ . Therefore there are as many  $C_K(X_k)$  as there are vertices of  $K$ , since the  $X_k$  are pairwise disjoint

and hence each carrier  $C_K(X_k)$  has only one vertex. Thus  $f^{-1}$  and hence  $f$  permute the vertices of  $K$ .

The family of sets,  $\{ \bigcup_{m \geq 0} \{f^m(k)\} : k \text{ is a vertex of } K \}$ , partition the vertices of  $K$  into  $\nu$  disjoint classes on each of which  $f$  is a cyclic permutation. Denote the convex hulls of these partitioning sets by  $K_1, K_2, \dots, K_\nu$ . Let  $k_{r0}$  be any vertex of  $K_r$ ; let  $k_{rs} = f^s(k_{r0})$  for  $r = 1, 2, \dots, \nu$  and  $s = 0, 1, 2, \dots, c_r - 1$ . Let  $C_{rs}$  denote the carrier in  $S$  of  $k_{rs}$ .

(c) Each  $C_{rs}$  meets  $K$  in only one point, namely  $k_{rs}$ , and hence  $C_{rs} = C_{r's'}$  iff  $(r, s) = (r', s')$ . If not  $C_{rs}$  would contain two distinct points  $k_{rs}$  and  $k'$  of  $K$ . The line they determine would meet  $C_{rs}$  in a line segment contained in  $K$  neither of whose endpoints is  $k_{rs}$ , contradicting the assumption that  $k_{rs}$  is a vertex.

(d)  $f(C_{rs}) \subseteq C_{r, s+1}$  and hence  $f^t(C_{rs}) \subseteq C_{rs'}$  if  $s' \equiv t + s \pmod{c_r}$ ; because, by remark (3),  $f(C_{rs})$  is contained in the carrier in  $S$  of  $f(k_{rs})$  which is  $C_{r, s+1}$  by definition.

$$(e) \bigcap_{m \equiv t} f^m(C_{rs}) = \{k_{rs'}\} \text{ if } s' \equiv t + s \pmod{c_r}.$$

To see why this is so we observe first that if  $m \equiv t \pmod{c_r}$ , then  $k_{rs'} \in f^m(C_{rs})$  because  $f^m(k_{rs}) = k_{r, s+m} = k_{rs'}$ ; and secondly that  $(\bigcap_{m \equiv t} f^m(C_{rs})) \subseteq K$  so that  $\{k_{rs'}\} \subseteq \bigcap_{m \equiv t} f^m(C_{rs}) = K \cap \bigcap_{m \equiv t} f^m(C_{rs}) \subseteq \bigcap_{m \equiv t} K \cap f^m(C_{rs}) \subseteq \bigcap_{m \equiv t} K \cap C_{rs'}$ .

According to (c),  $K \cap C_{rs'} = \{k_{rs'}\}$  and hence  $\bigcap_{m \equiv t} f^m(C_{rs}) = \{k_{rs'}\}$ . An immediate consequence of this is:

(f) The  $C_{rs}$  are pairwise disjoint. Applying remarks (2) and (1) we have:

(g)  $K$  is a simplex and the  $K_r$  are disjoint.

Evidently the set  $F$  of all fixed points is a convex subset of  $K$ . By linearity, the barycenter  $b_r = \frac{1}{c_r} \sum_{s=0}^{c_r-1} k_{rs}$  of  $K_r$

is fixed by  $f$  and hence  $F$  contains the convex hull of these barycenters. Conversely, if  $x$  is fixed, then, since  $x \in K$ ,

$$x = \sum_{r=1}^v \sum_{s=0}^{c_r-1} x_{rs} k_{rs} \quad (\text{where } x_{rs} \geq 0 \text{ and } \sum \sum x_{rs} = 1)$$

$$\text{and hence } f(x) = \sum_{r=1}^v \sum_{s=0}^{c_r-1} x_{rs} k_{r,s+1} = x. \quad \text{Therefore}$$

$x_{rs} = x_{rs'}$  where  $s' \equiv s + 1 \pmod{c_r}$  and hence, given  $r$ ,

either  $x_{rs} = 0$  for each  $0 \leq s < c_r$  or for all  $0 \leq s < c_r : x_{rs} = \frac{1}{c_r}$ .

Consequently  $x$  is in the convex hull of the barycenters.

Therefore  $F$  is a convex polytope spanned by the  $v$  barycenters of the  $K_r$ . The barycenters of the  $K_r$  are the vertices of  $F$

because the  $K_r$  are distinct. By applying remark (2) and (g)

we obtain:

(h) The set  $F$  of all  $f$ -fixed points is a simplex whose vertices are the barycenters of the  $K_r$ .

This completes the proof of the theorem.

We shall present a proof of the probabilistic form of the fundamental theorem (theorem I) after a few preliminary remarks showing the correspondence between the pertinent geometric and probabilistic ideas.

Each state  $i = 1, 2, \dots, n$  of the Markov chain whose one step transition matrix is  $A = (a_{ij})$  corresponds to the  $n$ -tuple  $v_i$  whose only non-zero component, 1, is its  $i$ -th component. Let  $f(x) = xA$  (i.e. the  $j$ -th component of  $f(x)$  is  $\sum_{i=1}^n x_i a_{ij}$ ) for each  $x$  in the convex hull  $S$ , of the  $v_i$ .  $S$  is a simplex and  $f$  is a linear operator on  $S$ . If  $X \subseteq S$  let  $C(X)$  denote the carrier of  $X$  in  $S$ . We then have:

$$(4) \quad i \rightsquigarrow j \text{ iff } v_j \in \bigcup_{m>0} C(f^m(v_i)) \text{ because of the definitions}$$

of  $\rightsquigarrow$  and  $C$ .

We shall show that

$$(5) \quad i \text{ is ergodic iff } v_i \text{ is a vertex of } C(K)$$

after we have established (5a) and (5b) below.

(5a) If  $E(K) = \{i : v_i \in C(K)\}$  then for each  $j$  there is an  $i \in E(K)$  such that  $j \rightsquigarrow i$ .

Proof of (5a). Let  $D = C(\{f^m(v_j) : m > 0\})$  then, using (3), we have  $f(D) \subseteq D$  and hence  $\bigcap_{m>0} f^m(D)$  is a non-empty subset of both  $K$  and  $D$ . There is, therefore, a vertex  $v_i$  of  $C(K)$  which is also a vertex of  $D$ . But the vertices of  $D$  are also those of  $\bigcup_{m>0} C(f^m(v_j))$ . Consequently  $j \rightsquigarrow i \in E(K)$ .

(5b) If  $i \in E(K)$  and  $i \rightsquigarrow j$  then  $j \in E(K)$  and  $j \rightsquigarrow i$ .

Proof of (5b).  $i \in E(K)$  and  $i \rightsquigarrow j$  imply that  $v_i$  is a vertex of some  $C_{rs}$  and  $v_j$  is a vertex of some  $C(f^t(v_i))$  by (4). Therefore  $f^t(v_i) \in f^t(C_{rs})$ . But  $f^t(C_{rs}) \subseteq C_{rs'}$  when  $s' \equiv s + t \pmod{c_r}$  according to Theorem 1 part (v) and hence

$v_j$  is a vertex of  $C_{rs'}$ . Consequently  $j \in E(K)$ .

$C(f_j^m(v_j)) = C_{rs}$  for a sufficiently large  $m \equiv s - s' \pmod{c_r}$  by parts (v) and (vi) of theorem 1, but  $v_i \in C_{rs}$  and hence  $j \rightsquigarrow i$  by (4).

Proof of (5).  $E(K) \subseteq E$  by (5b) and the definition of  $E$ . If  $j \notin E(K)$  then  $j \rightsquigarrow i \in E(K)$  by (5a) and  $i \not\rightsquigarrow j$  by (5b). Consequently  $j \notin E$  and hence  $E \subseteq E(K)$ .

### Proof of Theorem I.

(Ia) If  $i$  is any state let  $x^{(m)}$  be the point of  $C(K)$  closest to  $f_i^m(v_i)$ . Because of the definition of  $K$  the sequence of distances  $d_m$  between  $f_i^m(v_i)$  and  $x^{(m)}$  converges to zero.

According to (5)  $x_j^{(m)} = 0$  for all transient  $j$ ; therefore

$\lim_{m \rightarrow \infty} a_{ij}^{(m)} = 0$  because the  $j$ -th component of  $f_i^m(v_i)$  is  $a_{ij}^{(m)}$ .

Let  $E_{rs} = \{i : v_i \in C_{rs}\}$ ; then, evidently,

$E_r = \bigcup_{s=0}^{c_r-1} E_{rs}$ . The  $E_{rs}$  are pairwise disjoint and non-empty because of parts (iii) and (iv) of Theorem 1.

(Ib) If  $i \in E_{rs}$  and  $i$  leads to  $j$  in one step then  $v_i \in C_{rs}$  and  $v_j \in C(f(v_i))$ . But according to Theorem 1 part (v):  $f(v_i) \in C_{r,s+1}$  and hence  $v_j \in C_{r,s+1}$ .

Consequently  $j \in E_{r,s+1}$ .

(Ic)  $k_{rs}$  is an  $n$ -tuple of non-negative numbers summing to 1 because  $k_{rs} \in S$ . The  $j$ -th component,  $k_j^{(r,s)}$  of  $k_{rs}$  is 0 iff  $j \notin E_{rs}$  because of the definitions of  $C_{rs}$  and  $E_{rs}$ .

$$\lim_{m \rightarrow \infty} a_{ij}^{(mc_r + t)} = k_j^{(r, s')} \quad \text{for all } j \text{ and all}$$

$t = 0, 1, \dots, c_r - 1$  whenever  $i \in E_{rs}$  and  $s' \equiv s + t \pmod{c_r}$  by (vi).

This completes the proof of theorem I.

Theorem II can be proven either directly from Theorem I (their statements are equivalent) or from Theorem 1.

To obtain a proof of Theorem II the latter way let  $c(r, s)$  be the number of vertices in  $C_{rs}$  and let  $P$  be the permutation matrix which performs the change of basis mapping the first  $c$  basis vectors (vertices)  $v_1, v_2, \dots, v_c$  onto each of the  $c$  vertices not in  $C(K)$ , mapping the next  $c(1, 1)$  basis vectors  $v_{c+1}, v_{c+2}, \dots, v_{c+c(1, 1)}$  onto the vertices of  $C_{11}$ ; and so forth until all the last  $c(v, c_v - 1)$  basis vectors are mapped onto the vertices of  $C_{v, c_v - 1}$ .

(IIa) is proven analogously to (Ia). (IIb) is a result of (v).

To prove (IIc), define  $\pi_j^{(r, s)}$  by letting  $\pi_j^{(r, s)}$  be the  $j$ -th component of  $k_{rs}$  for each  $j = 1, 2, \dots, c(r, s)$ ; each  $s = 0, 1, \dots, c_r - 1$  and each  $r = 1, 2, \dots, v$ . (IIc) then follows from (vi).

We have extended the techniques used here to study inhomogeneous chains, that is, to study the asymptotic behaviour of products  $A_1 \cdot A_2 \cdot \dots \cdot A_n \cdot \dots$  of stochastic matrices  $A_n$  which are not necessarily the same matrix and all of which might be infinite. Some of these results are contained in a forthcoming paper on infinite products of substochastic matrices [7].

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