

HYPERMATRIX AND ITS APPLICATION

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In this paper, the author will consider multidimensional matrices on a line different from that of R. Gouarné and I. Samuel.¹ In the first half, a singular situation shall be pointed out concerning multidimensional matrices and in the second half, an application of a specific kind of them shall be proposed.

§0 Hypermatrices

A set out of numbers on the lattice points (i, j, \dots, k, l) of an N -dimensional space will be called an N -dimensional matrix of order $I \times J \times \dots \times K \times L$ or an $I \times J \times \dots \times K \times L$ hypermatrix, where i ranges over $1, 2, \dots, I$, j over $1, 2, \dots, J$, ..., k over $1, 2, \dots, K$ and l over $1, 2, \dots, L$.

The number $a_{i,j,\dots,k,l}$ set out on the point (i, j, \dots, k, l) will be called the $i \cdot j \cdot \dots \cdot k \cdot l$ element of the hypermatrix, which as a whole will be denoted by A or by $[a_{i,j,\dots,k,l}]$.

The elements shall be real numbers and the dimension be fixed hereafter.

Equality and inequality of two hypermatrices of the same order will be defined in the same way as usual. Thus they enjoy the usual fundamental properties: reflexivity, symmetricity or antisymmetricity and transitivity.

Addition of two hypermatrices of the same order will be defined also as usual, and it enjoys commutativity and associativity. There exists the additive identity and every hypermatrix has its additive reciprocal.

Scalar multiplication of a hypermatrix by a real number will be defined too as usual. And it enjoys commutativity, associativity and double distributivity.

Let $A = [a_{i,j,\dots,k,l}]$ be an $I \times J \times \dots \times K \times L$ hypermatrix and $B = [b_{p,q,\dots,r,s}]$ a $P \times Q \times \dots \times R \times S$ hypermatrix with $L = P$. Then multiplication of A by B shall be defined by

$$AB = \left[\sum_{l,q,\dots,r} a_{i,j,\dots,k,l} b_{l,q,\dots,r,s} \right].$$

Thus it proves easily to be associative and doubly distributive over addition.

§1 Cubic Hypermatrices

An $L \times L \times \dots \times L \times L$ hypermatrix will be called cubic. The set of all cubic hypermatrices with a specific L forms a ring with respect to addition and multiplication defined above, since it is closed under these operations which possess the fundamental properties as was shown

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¹ René Gouarné et Isaac Samuel: Introduction à l'étude des matrices multidimensionnelles. *Cahiers de Physique* 16 (1962), pp. 133-142.

Isaac Samuel et René Gouarné: Determinants des matrices à N dimension d'ordre n . *Cah. de Phys.* 16 (1962), pp. 143-152.

in the preceding section.

How about the multiplicative unit?

It might be natural to introduce the extended Kronecker's delta by

$$\delta_{ij \dots kl} = \begin{cases} 1 & \text{if } i=j=\dots=k=l \\ 0 & \text{otherwise} \end{cases}$$

Let the hypermatrix $[\delta_{ij \dots kl}]$ be denoted by E , then we have

$$(1) \quad AE = A$$

for every A . Because

$$\sum_{l, q, \dots, r} a_{ij \dots kl} \delta_{lq \dots rs} = a_{ij \dots ks}$$

holds for every N -ple (i, j, \dots, k, s) , where $a_{ij \dots kl}$ being the general element of A .

Though (1) is an identity in hypermatrix algebra,

$$(2) \quad EA = A$$

is not. Because

$$\sum_{l, q, \dots, r} \delta_{ij \dots kl} a_{lq \dots rs}$$

vanishes unless $i=j=\dots=k=l$.

It is a singular situation which does not occur in matrix algebra that (1) is an identity, but (2) is not. Thus E might be called justly a right unit.

How about the uniqueness of right unit?

For the sake of simplicity, two definitions shall be introduced.

Definition I: A cubic hypermatrix U is called a right unit, if

$$AU = A$$

holds for every cubic hypermatrix A .

Definition II: Let $A = [a_{ij \dots kl}]$ be a given cubic hypermatrix. The square matrix

$$[\sum_{j, \dots, k} a_{ij \dots kl}]$$

shall be called the contracted matrix of A and denoted by $\text{Ctr } A$.

Then we can prove

[Theorem 1] A cubic hypermatrix is a right unit if and only if its contracted matrix is the unit matrix.

Proof: If $U = [u_{ij \dots kl}]$ is a right unit, then the equality

$$(3) \quad \sum_{l, q, \dots, r} a_{ij \dots kl} u_{lq \dots rs} = a_{ij \dots ks}$$

i.e.

$$(4) \quad \sum_l (a_{ij \dots kl} \sum_{q, \dots, r} u_{lq \dots rs}) = a_{ij \dots ks}$$

holds for every $a_{ij \dots ks}$.

Therefore

$$(5) \quad \sum_{q, \dots, r} u_{lq \dots rs} = \delta_{ls}$$

i.e.

$$(6) \quad \text{Ctr } U = E^{(2)},$$

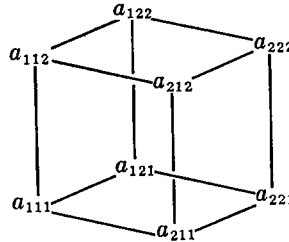
where $E^{(2)}$ is the unit matrix. Thus the condition is necessary.

Conversely, let (6) hold. Then we have elementwise (5), which implies (4) i.e. (3) for every $a_{ij \dots ks}$. Thus the condition is sufficient, and the proof is completed.

This theorem tells us that right unit is not at all unique. Because L^2 equations (5) have L^N unknowns.

And this is another singular situation concerning hypermatrices.

A $2 \times 2 \times 2$ hypermatrix



will be denoted by

$$\begin{bmatrix} a_{111} & a_{121} & a_{112} & a_{122} \\ a_{211} & a_{221} & a_{212} & a_{222} \end{bmatrix}$$

for the sake of typographical simplicity.

Example 1 Two right units shall be given for instance.

$$\begin{bmatrix} 3 & 4 & -5 & 4 \\ 2 & 3 & 9 & -7 \end{bmatrix} \begin{bmatrix} 0.5 & 0.5 \\ -1 & 1 \end{bmatrix} \begin{bmatrix} 2 & -2 \\ -3 & 4 \end{bmatrix} = \begin{bmatrix} 3 & 4 & -5 & 4 \\ 2 & 3 & 9 & -7 \end{bmatrix},$$

$$\begin{bmatrix} 3 & 4 & -5 & 4 \\ 2 & 3 & 9 & -7 \end{bmatrix} \begin{bmatrix} 4 & -3 \\ -2 & 2 \end{bmatrix} \begin{bmatrix} 1 & -1 \\ 0.5 & 0.5 \end{bmatrix} = \begin{bmatrix} 3 & 4 & -5 & 4 \\ 2 & 3 & 9 & -7 \end{bmatrix}.$$

Concerning contracted matrices, we have

[Lemma] The contracted matrix of the product of two cubic hypermatrices is the product of their contracted matrices.

Proof: Let $A = [a_{ij...kl}]$ and $B = [b_{pq...rs}]$ be two given cubic hypermatrices.

Then by definition

$$\begin{aligned} \text{Ctr}(AB) &= \text{Ctr} \left[\sum_{l,q,\dots,r} a_{ij...kl} b_{lq...rs} \right] \\ &= \left[\sum_{j,\dots,k} \left(\sum_{l,q,\dots,r} a_{ij...kl} b_{lq...rs} \right) \right] \\ &= \left[\sum_l \left(\sum_{j,\dots,k} a_{ij...kl} \right) \left(\sum_{q,\dots,r} b_{lq...rs} \right) \right] \\ &= \left[\sum_{j,\dots,k} a_{ij...kl} \right] \left[\sum_{q,\dots,r} b_{lq...rs} \right]. \end{aligned}$$

The last member is just the product of the contracted matrix of A and that of B . And the lemma is proved.

How about the reciprocal?

For the sake of this problem, let us introduce

Definition III: Let A be a given cubic hypermatrix and U a right unit. If there exists a cubic hypermatrix B such that $AB=U$, then A is said to be nonsingular and B is called a right reciprocal of A .

And we have

[Theorem 2] A cubic hypermatrix is nonsingular if and only if its contracted matrix is nonsingular.

Proof: A cubic hypermatrix A is nonsingular if and only if there exists a cubic hypermatrix B such that

$$(7) \quad AB = U,$$

where U is a right unit. On applying the lemma we have

$$\text{Ctr } A \cdot \text{Ctr } B = \text{Ctr } U$$

which combines Theorem 1 to yield

$$(8) \quad \text{Ctr } A \cdot \text{Ctr } B = E^{(2)}.$$

Thus the nonsingularity of $\text{Ctr } A$ is a necessary condition.

Conversely, let $\text{Ctr } A$ be a nonsingular matrix. By definition there exists a matrix M such that

$$(\text{Ctr } A)M = E^{(2)}.$$

Then we can construct easily a cubic hypermatrix B , whose contracted matrix is M . And we have (8). Application of the lemma on the left member yields

$$\text{Ctr } (AB) = E^{(2)}.$$

which combines Theorem 1 to yield (7).

Thus the condition is sufficient and the proof is completed.

(Corollary) If AB is a right unit, then so is BA .

Proof: (8) and commutativity of reciprocal matrices together imply

$$\text{Ctr } B \cdot \text{Ctr } A = E^{(2)}$$

which is equivalent with

$$BA = U$$

by Theorem 1.

Example 2

$$\left[\begin{array}{cc|cc} 2 & 1 & 3 & 1 \\ -1 & 3 & 2 & 1 \end{array} \right]$$

is nonsingular because of

$$\left[\begin{array}{cc|cc} 2 & 1 & 3 & 1 \\ -1 & 3 & 2 & 1 \end{array} \right] \left[\begin{array}{cc|cc} 1 & 2 & -3 & -1 \\ 1 & -3 & 2 & 1 \end{array} \right] = \left[\begin{array}{cc|cc} 0 & 1 & 1 & -1 \\ -7 & 7 & 10 & -9 \end{array} \right]$$

and

$$\text{Ctr} \left[\begin{array}{cc|cc} 0 & 1 & 1 & -1 \\ -7 & 7 & 10 & -9 \end{array} \right] = \left[\begin{array}{cc} 1 & 0 \\ 0 & 1 \end{array} \right].$$

$$\left[\begin{array}{cc|cc} 1 & 2 & -3 & -1 \\ 1 & -3 & 2 & 1 \end{array} \right] \left[\begin{array}{cc|cc} 2 & 1 & 3 & 1 \\ -1 & 3 & 2 & 1 \end{array} \right] = \left[\begin{array}{cc|cc} -3 & 4 & -5 & 5 \\ 7 & -7 & 10 & -9 \end{array} \right]$$

and

$$\text{Ctr} \left[\begin{array}{cc|cc} -3 & 4 & -5 & 5 \\ 7 & -7 & 10 & -9 \end{array} \right] = \left[\begin{array}{cc} 1 & 0 \\ 0 & 1 \end{array} \right]$$

combine the first half to verify the lemma.

Theorem 2 and the corollary are very close to those of matrix algebra except the uniqueness of reciprocal.

§2 Transition Hypermatrices

Let us consider a specific kind of hypermatrices in this section.

Definition IV: A cubic hypermatrix $A = [a_{ij..kl}]$ will be called a transition hypermatrix if $A \geq O$ and

$$\sum_l a_{ij..kl} = 1 \quad \text{for all } (i, j, \dots, k),$$

where O is a cubic hypermatrix with zero elements exclusively.

In order to apply this kind of hypermatrices in somewhat practical situation, the following definition shall be introduced.

Definition V: Let $A=[a_{ij...kl}]$ and $B=[b_{ij...kl}]$ be transition hypermatrices of order L^N . Then the hypermatrix

$$\left[\frac{1}{L^{N-2}} \sum_{l,q,...,r} a_{ij...kl} b_{lq...rs} \right],$$

will be called the transition product of A and B and denoted by $A*B$.

[Theorem 3] The transition product of two transition hypermatrices is a transition hypermatrix.

Proof: Let A and B be given transition hypermatrices of order L^N .

Due to nonnegativity of the elements and Definition V, we have

$$(1) \quad \begin{aligned} A*B &\geq O, \\ \sum_l a_{ij...kl} &= 1 \quad \text{for all } (i, j, \dots, k) \end{aligned}$$

and

$$\sum b_{lq...rs} = 1 \quad \text{for all } (l, q, \dots, r)$$

together imply that

$$\begin{aligned} \sum_s \left(\frac{1}{L^{N-2}} \sum_{l,q,...,r} a_{ij...kl} b_{lq...rs} \right) \\ = \frac{1}{L^{N-2}} \sum_{q,...,r} \left(\sum_l a_{ij...kl} \sum_s b_{lq...rs} \right) \\ = \frac{1}{L^{N-2}} \sum_{q,...,r} 1. \end{aligned}$$

Since each of the $N-2$ numbers q, \dots, r ranges over $1, 2, \dots, L$,
 $\sum_{q,...,r} 1 = L^{N-2}$.

Thus we have

$$(2) \quad \sum_s \left(\frac{1}{L^{N-2}} \sum_{l,q,...,r} a_{ij...kl} b_{lq...rs} \right) = 1.$$

(1) and (2) combine to prove that $A*B$ is also a transition hypermatrix.

Again for the sake of application, let us introduce a specific kind of transition hypermatrices by a usual

Definition VI: A transition hypermatrix A will be said to be regular, if there exists a natural number H such that

$${}^H A > O,$$

where presuperscript denotes a transition power of a transition hypermatrix.

And we can prove the

[Principal Theorem 1] If A is a regular transition hypermatrix, then

$${}^n A \rightarrow W \quad \text{as } n \rightarrow \infty,$$

where $W=[w_{ij...kl}]$ is a transition hypermatrix with such elements as

$$w_{ij...kl} = w_l$$

irrespective of (i, j, \dots, k) .

Proof: We can assume without loss of generality that A is positive. Because otherwise, there exists a natural number H with ${}^H A > O$, and we can prove that

$${}^m({}^H A) \rightarrow W \quad \text{as } m \rightarrow \infty$$

i.e.

$${}^n A \rightarrow W \quad \text{as } n \rightarrow \infty.$$

Let μ be defined by

$$\mu = \min_{i,j,\dots,k,l} a_{ij\dots kl},$$

then due to the assumption and Definition IV the inequalities

$$(3) \quad 0 < \mu < 1$$

hold.

s shall be specified hereafter and M_t, m_t and $a_{ij\dots ks}^{(t)}$ defined by induction with respect to t .

We will begin with

$$(4) \quad M_0 = \max_{i,j,\dots,k} a_{ij\dots ks}$$

$$(5) \quad m_0 = \min_{i,j,\dots,k} a_{ij\dots ks},$$

$$(6) \quad a_{ij\dots ks}^{(1)} = \frac{1}{L^{N-2}} \sum_{l,q,\dots,r} a_{ij\dots kl} a_{lq\dots rs},$$

$$(7) \quad M_1 = \max_{i,j,\dots,k} a_{ij\dots ks}^{(1)},$$

and

$$(8) \quad m_1 = \min_{i,j,\dots,k} a_{ij\dots ks}^{(1)}.$$

We have

$$(9) \quad m_0 \leq M_0 \quad \text{and} \quad m_1 \leq M_1$$

obviously by definitions.

On substituting M_0 for $a_{lq\dots rs}$ in (6) and simplifying by $\sum_l a_{ij\dots kl} = 1$, we obtain

$$a_{ij\dots ks}^{(1)} \leq M_0$$

which combines (7) to prove

$$(10) \quad M_1 \leq M_0.$$

Similarly, substitution of m_0 for $a_{lq\dots rs}$ in (6) and simplification by $\sum_l a_{ij\dots kl} = 1$ yield

$$a_{ij\dots ks}^{(1)} \geq m_0$$

which is combined with (8) to have

$$(11) \quad m_1 \geq m_0.$$

Due to (9), (10) and (11), inequalities

$$(12) \quad m_0 \leq m_1 \leq M_1 \leq M_0$$

hold.

Now we will define $a_{ij\dots ks}^{(t)}$, M_t and m_t by induction

$$(13) \quad a_{ij\dots ks}^{(t)} = \frac{1}{L^{N-2}} \sum_{l,q,\dots,r} a_{ij\dots kl} a_{lq\dots rs}^{(t-1)},$$

$$(14) \quad M_t = \max_{i,j,\dots,k} a_{ij\dots ks}^{(t)},$$

$$(15) \quad m_t = \min_{i,j,\dots,k} a_{ij\dots ks}^{(t)}.$$

The inequalities (12) mean that inequalities

$$(16) \quad m_0 \leq m_1 \leq \dots \leq m_{t-1} \leq M_{t-1} \leq \dots \leq M_1 \leq M_0$$

hold for $t=2$.

By substituting M_{t-1} for $a_{lq\dots rs}^{(t-1)}$ in (13) and simplifying by $\sum_l a_{ij\dots kl} = 1$, we have

$$a_{ij..ks}^{(l)} \leq M_{t-1}$$

which combines (14) to yield

$$(17) \quad M_t \leq M_{t-1}.$$

Similarly, on substituting m_{t-1} for $a_{lq..rs}^{(l-1)}$ in (13) and applying $\sum_l a_{ij...kl} = 1$, it is observed that

$$a_{ij..ks}^{(l)} \geq m_{t-1}.$$

And we have

$$(18) \quad m_t \geq m_{t-1}$$

by definition (15).

(16), (17) and (18) together imply that

$$(19) \quad m_0 \leq m_1 \leq \dots \leq m_t \leq M_t \leq \dots \leq M_1 \leq M_0.$$

Thus (19) has proved to be true for all t .

By definitions, two classes of inequalities

$$a_{ij..kl} - \mu \geq 0 \quad \text{for all } (i, j, \dots, k, l)$$

and

$$M_0 - a_{lq..rs} \geq 0 \quad \text{for all } (l, q, \dots, r)$$

hold. Therefore we have

$$(a_{ij..kl} - \mu)(M_0 - a_{lq..rs}) \geq 0$$

i.e.

$$a_{ij..kl}M_0 - \mu M_0 + \mu a_{lq..rs} \geq a_{ij..kl}a_{lq..rs}$$

for all (i, j, \dots, k, l) and all (l, q, \dots, r) .

Summation over (l, q, \dots, r) and division by L^{N-2} yield

$$\frac{M_0}{L^{N-2}} \sum_{l,q,\dots,r} a_{ij..kl} - L\mu M_0 + \frac{\mu}{L^{N-2}} \sum_{l,q,\dots,r} a_{lq..rs} \geq \frac{1}{L^{N-2}} \sum_{l,q,\dots,r} a_{ij..kl}a_{lq..rs}.$$

On applying $\sum_l a_{ij...kl} = 1$ and (6), we obtain

$$M_0 - L\mu M_0 + \frac{\mu}{L^{N-2}} \sum_{l,q,\dots,r} a_{lq..rs} \geq a_{ij..ks}^{(1)} \quad \text{for all } (i, j, \dots, k).$$

Thus by (7), the inequality

$$(20) \quad (1 - L\mu)M_0 + \frac{\mu}{L^{N-2}} \sum_{l,q,\dots,r} a_{lq..rs} \geq M_1$$

holds.

Similarly by definitions, we have two classes of inequalities

$$a_{ij..kl} - \mu \geq 0 \quad \text{for all } (i, j, \dots, k, l)$$

and

$$a_{lq..rs} - m_0 \geq 0 \quad \text{for all } (l, q, \dots, r),$$

whence we infer that

$$(a_{ij..kl} - \mu)(a_{lq..rs} - m_0) \geq 0,$$

i.e.

$$a_{ij..kl}m_0 - \mu m_0 + \mu a_{lq..rs} \leq a_{ij..kl}a_{lq..rs} \quad \text{for all } (i, j, \dots, k, l) \quad \text{and all } (l, q, \dots, r).$$

We sum up these inequalities over (l, q, \dots, r) and divide by L^{N-2} to obtain

$$\frac{m_0}{L^{N-2}} \sum_{l,q,\dots,r} a_{ij..kl} - L\mu m_0 + \frac{\mu}{L^{N-2}} \sum_{l,q,\dots,r} a_{lq..rs} \leq \frac{1}{L^{N-2}} \sum_{l,q,\dots,r} a_{ij..kl}a_{lq..rs}.$$

Application of $\sum_l a_{ij...kl} = 1$ and (6) yield

$$m_0 - L\mu m_0 + \frac{\mu}{L^{N-2}} \sum_{l, q, \dots, r} a_{lq \dots rs} \leq a_{ij \dots ks}^{(1)} \quad \text{for all } (i, j, \dots, k).$$

Thus by (8) the inequality

$$(21) \quad (1 - L\mu)m_0 + \frac{\mu}{L^{N-2}} \sum_{l, q, \dots, r} a_{lq \dots rs} \leq m_1$$

holds.

Subtracting (21) from (20), we obtain

$$(1 - L\mu)(M_0 - m_0) \geq M_1 - m_1$$

which concludes

$$1 - L\mu \geq 0$$

and that

$$(22) \quad (1 - L\mu)^t (M_0 - m_0) \geq M_t - m_t,$$

is true for $t=1$.

Let us assume that (22) is true for $t=u-1$ i.e.

$$(23) \quad (1 - L\mu)^{u-1} (M_0 - m_0) \geq M_{u-1} - m_{u-1}.$$

By definitions, we have two classes of inequalities

$$a_{ij \dots kl} - \mu \geq 0 \quad \text{for all } (i, j, \dots, k, l)$$

and

$$M_{u-1} - a_{lq \dots rs}^{(u-1)} \geq 0 \quad \text{for all } (l, q, \dots, r),$$

whence we obtain

$$(a_{ij \dots kl} - \mu)(M_{u-1} - a_{lq \dots rs}^{(u-1)}) \geq 0$$

i.e.

$$a_{ij \dots kl} M_{u-1} - \mu M_{u-1} + \mu a_{lq \dots rs}^{(u-1)} \geq a_{ij \dots kl} a_{lq \dots rs}^{(u-1)} \quad \text{for all } (i, j, \dots, k, l) \text{ and all } (l, q, \dots, r).$$

On summing up these inequalities over (l, q, \dots, r) and dividing by L^{N-2} , we have

$$\frac{M_{u-1}}{L^{N-2}} \sum_{l, q, \dots, r} a_{ij \dots kl} - L\mu M_{u-1} + \frac{\mu}{L^{N-2}} \sum_{l, q, \dots, r} a_{lq \dots rs}^{(u-1)} \geq \frac{1}{L^{N-2}} \sum_{l, q, \dots, r} a_{ij \dots kl} a_{lq \dots rs}^{(u-1)},$$

which is combined with $\sum_{l, q, \dots, r} a_{ij \dots kl} = 1$, (13) and (14) to yield

$$(24) \quad (1 - L\mu)M_{u-1} + \frac{\mu}{L^{N-2}} \sum_{l, q, \dots, r} a_{lq \dots rs}^{(u-1)} \geq M_u.$$

Similarly, we have two classes of inequalities

$$a_{ij \dots kl} - \mu \geq 0 \quad \text{for all } (i, j, \dots, k, l)$$

and

$$a_{lq \dots rs}^{(u-1)} - m_{u-1} \geq 0 \quad \text{for all } (l, q, \dots, r)$$

by definitions.

We combine them to obtain

$$(a_{ij \dots kl} - \mu)(a_{lq \dots rs}^{(u-1)} - m_{u-1}) \geq 0$$

i.e.

$$a_{ij \dots kl} m_{u-1} - \mu m_{u-1} + \mu a_{lq \dots rs}^{(u-1)} \leq a_{ij \dots kl} a_{lq \dots rs}^{(u-1)} \quad \text{for all } (i, j, \dots, k, l) \text{ and all } (l, q, \dots, r).$$

Summation over (l, q, \dots, r) and division by L^{N-2} yield

$$\frac{m_{u-1}}{L^{N-2}} \sum_{l, q, \dots, r} a_{ij \dots kl} - L\mu m_{u-1} + \frac{\mu}{L^{N-2}} \sum_{l, q, \dots, r} a_{lq \dots rs}^{(u-1)} \leq \frac{1}{L^{N-2}} \sum_{l, q, \dots, r} a_{ij \dots kl} a_{lq \dots rs}^{(u-1)}$$

which we simplify into

$$(25) \quad (1-L_\mu)m_{u-1} + \frac{\mu}{L^{N-2}} \sum_{l,q,\dots,r} a_{lq}^{(u-1)} \leq m_u$$

by $\sum_l a_{ij\dots kl}=1$, (13) and (15).

Subtraction of (25) from (24) yields

$$(26) \quad (1-L_\mu)(M_{u-1}-m_{u-1}) \geq M_u - m_u.$$

(23) and (26) together with $(1-L_\mu) > 0$ imply that

$$(1-L_\mu)^u(M_0-m_0) \geq M_u - m_u$$

i.e. (22) is true for $t=u$.

Thus the general validity of (22) is proved by induction.

From (1) and (22) we can infer that

$$(27) \quad M_t - m_t \rightarrow 0 \quad \text{as } t \rightarrow \infty,$$

since $0 \leq 1-L_\mu < 1$.

(19) and (27) together conclude that $a_{ij\dots kl}^{(t)}$ is the $i \cdot j \cdot \dots \cdot k \cdot l$ element of tA , and there exists a number w_l for each l such that

$$a_{ij\dots kl}^{(t)} \rightarrow w_l \quad \text{as } t \rightarrow \infty,$$

independent of (i, j, \dots, k) .

And this is what is to be proved.

§3 An Application

Let us assume that there are L states given and from every $N-1$ ple combination of L states one of the L states may result. If the probability that the state l may result from a state-combination (i, j, \dots, k) is $a_{ij\dots kl}$ with $\sum_l a_{ij\dots kl}=1$, then this situation might be realised compactly by the L^N transition hypermatrix $[a_{ij\dots kl}]$.

Example 3 The following table gives the percentage of children with bright-coloured hair.

FATHER \ MOTHER	BRIGHT	DARK
	BRIGHT	DARK
BRIGHT	80	50
DARK	30	60

This is realised by 2^3 transition hypermatrix

$$\left[\begin{array}{cc|cc} 0.8 & 0.5 & 0.2 & 0.5 \\ 0.3 & 0.6 & 0.7 & 0.4 \end{array} \right].$$

The following theorem is very important in this connection.

[Theorem 4] The i, j, \dots, k, l element of nA gives the probability that the process which has started in a state-combination (i, j, \dots, k) will be in a state l after n steps provided that every state-combination is equally likely.

Proof: The theorem is true for $n=1$ by the above-given interpretation of a transition hypermatrix.

Assume that the theorem is true for $n=m-1$, and let rA be denoted by $[a_{ij\dots kl}^{(r)}]$.

The $i \cdot j \cdot \dots \cdot k \cdot l$ element of ${}^m A$ is

$$(1) \quad a_{ij\,kl}^{(m)} = \frac{1}{L^{N-2}} \sum_{p,q,r} a_{ij\,kp}^{(m-1)} a_{pq\,rl},$$

Since by assumption $a_{ij\,kp}^{(m-1)}$ is the probability that the process starting in a state-combination (i, j, \dots, k) will be in a state p after $m-1$ steps, $1/L^{N-2}$ is that of p to be combined with (q, \dots, r) and $a_{pq\,rl}$ is that of (p, q, \dots, r) to be followed by a state l on the m -th step, each summand of (1) is the probability that the process which started in a state-combination (i, j, \dots, k) will be in state l through a state-combination (p, q, \dots, r) after m steps. Such summands being summed up over (p, q, \dots, r) give the probability that the process which has started in a state-combination (i, j, \dots, k) will be in a state l after m steps. Thus the theorem is true for $n=m$.

And the proof is completed by induction.

Example 4 (Continued)

$$^2 \begin{bmatrix} 0.8 & 0.5 & | & 0.2 & 0.5 \\ 0.3 & 0.6 & | & 0.7 & 0.4 \end{bmatrix} = \begin{bmatrix} 0.61 & 0.55 & | & 0.39 & 0.45 \\ 0.51 & 0.57 & | & 0.49 & 0.43 \end{bmatrix},$$

tells that 55% of the grandchildren of bright-haired fathers and dark-haired mothers may be bright-haired.

Implication of Principal Theorem 1 is now very important, since it tells that after a large number of steps, the probability of the process to be in a specific state l will be nearly w_l no matter what the initial state-combination may be.

This situation shall be explained by

Example 5 (Continued)

$$\begin{aligned} ^3 \begin{bmatrix} 0.8 & 0.5 & | & 0.2 & 0.5 \\ 0.3 & 0.6 & | & 0.7 & 0.4 \end{bmatrix} &= \begin{bmatrix} 0.572 & 0.56 & | & 0.428 & 0.44 \\ 0.552 & 0.564 & | & 0.448 & 0.436 \end{bmatrix}, \\ ^4 \begin{bmatrix} 0.8 & 0.5 & | & 0.2 & 0.5 \\ 0.3 & 0.6 & | & 0.7 & 0.4 \end{bmatrix} &= \begin{bmatrix} 0.5644 & 0.562 & | & 0.4356 & 0.438 \\ 0.5604 & 0.5628 & | & 0.4396 & 0.4372 \end{bmatrix}, \\ ^5 \begin{bmatrix} 0.8 & 0.5 & | & 0.2 & 0.5 \\ 0.3 & 0.6 & | & 0.7 & 0.4 \end{bmatrix} &= \begin{bmatrix} 0.56288 & 0.5624 & | & 0.43712 & 0.4376 \\ 0.56208 & 0.5626 & | & 0.43792 & 0.4374 \end{bmatrix}. \end{aligned}$$

The next problem to be attacked is

how to get W in Principal Theorem 1.

Definition VII Let A be a given transition hypermatrix. A transition hypermatrix F for which

$$F * A = F,$$

holds will be called a fixed transition hypermatrix of A .

Now we can prove

[Principal Theorem 2] W in Principal Theorem 1 is the unique fixed transition hypermatrix of A .

Proof: By Principal Theorem 1, we have

$$(2) \quad {}^n A \rightarrow W \quad \text{as } n \rightarrow \infty.$$

Whence we obtain

$$(3) \quad {}^{n+1} A \rightarrow W * A \quad \text{as } n \rightarrow \infty.$$

(2) combined with (3) yields

$$W * A - W \rightarrow 0.$$

Since two hypermatrices in the left member are constant hypermatrices, we have

$$W * A = W,$$

which means that W is a fixed transition hypermatrix of A .

Let F be a fixed transition hypermatrix, then we have

$$F * A = F.$$

By successive postmultiplication of A we obtain

$$F *^n A = F,$$

which combines (2) to yield

$$F \rightarrow F * W.$$

Since both members are constant, we infer that

$$F = F * W$$

i.e.

$$f_{ij...kl} = \frac{1}{L^{N-2}} \sum_{p,q,\dots,r} f_{ij...kp} w_{pq...rl} \quad \text{for all } (i, j, \dots, k, l).$$

On applying $w_{pq...rl} = w_l$, it is observed that

$$\frac{1}{L^{N-2}} \sum_{p,q,\dots,r} f_{ij...kp} w_{pq...rl} = \frac{w_l}{L^{N-2}} \sum_{p,q,\dots,r} f_{ij...kp}.$$

Simplification by $\sum_p f_{ij...kp} = 1$ yields

$$\frac{w_l}{L^{N-2}} \sum_{p,q,\dots,r} f_{ij...kp} = w_l.$$

These three equalities combine to prove

$$f_{ij...kl} = w_l \quad \text{for all } (i, j, \dots, k, l)$$

which is equivalent with

$$F = W.$$

Thus the uniqueness is proved to complete the proof.

By Principal Theorem 2, we can evaluate the ultimate distribution of states, as shown in Example 6 (Continued)

$$\begin{bmatrix} w_1 & w_1 | w_2 & w_2 \\ w_1 & w_1 | w_2 & w_2 \end{bmatrix} * \begin{bmatrix} 0.8 & 0.5 | 0.2 & 0.5 \\ 0.3 & 0.6 | 0.7 & 0.4 \end{bmatrix} = \begin{bmatrix} w_1 & w_1 | w_2 & w_2 \\ w_1 & w_1 | w_2 & w_2 \end{bmatrix}$$

gives

$$\begin{cases} 0.65w_1 + 0.45w_2 = w_1 \\ 0.35w_1 + 0.55w_2 = w_2 \\ w_1 + w_2 = 1 \end{cases}$$

And we have

$$w_1 = 0.5625, \quad w_2 = 0.4375$$

which means that

$${}^n \begin{bmatrix} 0.8 & 0.5 | 0.2 & 0.5 \\ 0.3 & 0.6 | 0.7 & 0.4 \end{bmatrix} \rightarrow \begin{bmatrix} 0.5625 & 0.5625 | 0.4375 & 0.4375 \\ 0.5625 & 0.5625 | 0.4375 & 0.4375 \end{bmatrix} \quad \text{as } n \rightarrow \infty.$$

§4 Concluding Remarks

Thus far, we have considered the concept "hypermatrix" from applicative point of view on Markov chain. And some relevant results have been reached as far as regular Markov chain concerns. In §1, however, some singular situations were referred to concerning unit and reciprocal hypermatrices. Therefore we are little ready for absorbing Markov chain. The next problems are to elaborate definitions and to arrange theorems thereof.