## A RELATIONSHIP BETWEEN ARBITRARY POSITIVE MATRICES AND STOCHASTIC MATRICES

RICHARD SINKHORN

1. Introduction. The author (2) has shown that corresponding to each positive square matrix $A$ (i.e. every $a_{i j}>0$ ) is a unique doubly stochastic matrix of the form $D_{1} A D_{2}$, where the $D_{i}$ are diagonal matrices with positive diagonals. This doubly stochastic matrix can be obtained as the limit of the iteration defined by alternately normalizing the rows and columns of $A$.

In this paper, it is shown that with a sacrifice of one diagonal $D$ it is still possible to obtain a stochastic matrix. Of course, it is necessary to modify the iteration somewhat. More precisely, it is shown that corresponding to each positive square matrix $A$ is a unique stochastic matrix of the form $D A D$ where $D$ is a diagonal matrix with a positive diagonal. It is shown further how this stochastic matrix can be obtained as a limit to an iteration on $A$.

Immediate corollaries to this result are a theorem of Marcus and Newman (1), which states that if $A$ is a positive symmetric matrix, then there exists a diagonal matrix $D$ with a positive main diagonal such that $D A D$ is doubly stochastic, and its generalization, which states that if $A$ is positive $N \times N$ and if $p_{1}, \ldots, p_{N}$ are positive real numbers, then there exists a unique matrix of the form $D A D$ with row sums $p_{1}, \ldots, p_{N}$ where $D$ is a diagonal matrix with a positive diagonal.
2. Stochastic matrices and positive matrices. The main result is:

Theorem. Corresponding to each positive matrix $A$ there exists a unique stochastic matrix of the form $D A D$ where $D$ is a diagonal matrix with a positive diagonal.

The existence part of the proof is absorbed into three lemmas which follow.
Lemma 1. Let $V \subseteq E^{N} \times E^{N}$ consist of vector pairs $(x, y)$ with positive components that satisfy

$$
\sum_{j=1}^{N} y_{i} a_{i j} x_{j}=1, \quad i=1, \ldots, N
$$

with $\|x\|=\max \left|x_{i}\right| \leqslant a^{-\frac{1}{2}}$ and $\|y\|=\max \left|y_{i}\right| \leqslant a^{-\frac{1}{2}}$ where $a$ is the minimal element of the positive matrix $A=\left(a_{i j}\right)$. Then the function

$$
\phi(x, y)=\max _{i} \sum_{j=1}^{N} x_{i} a_{i j} y_{i}-\min _{i} \sum_{j=1}^{N} x_{i} a_{i j} y_{j}
$$

achieves a minimum of zero on $V$.
Received November 30, 1964.

Proof. Certainly $V$ is not empty since it contains $\left(x^{0}, y^{0}\right)$ where

$$
x_{i, 0}=a^{-\frac{1}{2}}, \quad y_{i, 0}=\left(\sum_{j} a_{i j}\right)^{-1} a^{\frac{1}{2}}, \quad \text { for } i=1, \ldots, N .
$$

Note that

$$
\left|y_{i, 0}\right| \leqslant a^{\frac{1}{2}} / a_{i j} \leqslant a^{\frac{1}{2}} / a=a^{-\frac{1}{2}}
$$

for any $i, j$.
Construct a sequence $\left(x^{n}, y^{n}\right) \in V$ as follows. Let $\left(x^{0}, y^{0}\right)$ be as above and set

$$
x_{i, n+1}=M_{n}^{-1} a^{-\frac{1}{2}} \rho_{i, n}^{-1} x_{i, n}, \quad y_{j, n+1}=M_{n} a^{\frac{1}{2}} \delta_{j, n}^{-1} y_{j, n},
$$

where

$$
\begin{aligned}
& \rho_{i, n}=\sum_{j} x_{i, n} a_{i j} y_{j, n}, \quad \delta_{j, n}=\sum_{i} \rho_{i, n}^{-1} x_{i, n} a_{j i} y_{j, n}, \\
& M_{n}=\max _{i} \rho_{i, n}^{-1} x_{i, n} .
\end{aligned}
$$

It is easy to see that each $\left(x^{n}, y^{n}\right)$ lies in $V$, for certainly $\sum_{j} y_{i, n} a_{i j} x_{j, n}=1$ for all $i$. Since for all $i, j, n$,

$$
\delta_{j, n}^{-1} y_{j, n}=\left(\sum_{i} \rho_{i, n}^{-1} x_{i, n} a_{j i}\right)^{-1} \leqslant\left(\rho_{i, n}^{-1} x_{i, n} a_{j i}\right)^{-1} \leqslant a^{-1}\left(\rho_{i, n}^{-1} x_{i, n}\right)^{-1}
$$

in particular

$$
\delta_{j, n}^{-1} y_{j, n} \leqslant a^{-1} M_{n}^{-1}
$$

for all $j$ and $n$. Thus

$$
y_{j, n+1} \leqslant M_{n} a^{\frac{1}{2}} a^{-1} M_{n}^{-1}=a^{-\frac{1}{2}} ;
$$

also

$$
x_{i, n+1} \leqslant M_{n}^{-1} a^{-\frac{1}{2}} M_{n}=a^{-\frac{1}{2}},
$$

and hence

$$
\left\|x^{n}\right\| \leqslant a^{-\frac{1}{2}} \quad \text { and } \quad\left\|y^{n}\right\| \leqslant a^{-\frac{1}{2}} \quad \text { for all } n
$$

Then from $x_{i, n} \sum_{j} a_{i j} y_{j, n}=\rho_{i, n}$, it follows that

$$
\rho_{i, n}^{-1} x_{i, n}=\left(\sum_{j} a_{i j} y_{j, n}\right)^{-1} \geqslant a^{\frac{1}{2}}\left(\sum_{j} a_{i j}\right)^{-1} \geqslant R a^{\frac{1}{2}},
$$

where $R^{-1}=\max _{i} \sum_{j} a_{i j}$. Also

$$
y_{j, n}=\left(\sum_{i} x_{i, n} a_{j i}\right)^{-1} \geqslant a^{\frac{1}{2}}\left(\sum_{i} a_{j i}\right)^{-1} \geqslant R a^{\frac{1}{2}},
$$

and therefore, in particular,

$$
d_{n}=\min _{i, j} \rho_{i, n}^{-1} x_{i, n} a_{i j} y_{j, n} \geqslant R a^{\frac{1}{2}} a R a^{\frac{1}{2}}=R^{2} a^{2}=\mu>0 \quad \text { for all } n .
$$

Let

$$
\begin{aligned}
\rho_{i_{1}, n+1}=\min _{i} \rho_{i, n+1}, & \rho_{i_{2}, n+1}=\max _{i} \rho_{i, n+1}, \\
\delta_{j_{1}, n}=\min _{j} \delta_{j, n}, & \delta_{j_{2}, n}=\max _{j} \delta_{j, n} .
\end{aligned}
$$

Then

$$
\begin{aligned}
\phi\left(x^{n+1}, y^{n+1}\right) & =\rho_{i_{2}, n+1}-\rho_{i_{1}, n+1} \\
& \leqslant\left[x_{i_{2}, n}^{-1} \rho_{i_{2}, n} a_{i_{2}, j_{2}} \delta_{j_{2}, n}^{-1} y_{j_{2}, n}+\delta_{j_{1}, n}^{-1}\left(1-x_{i_{2}, n} \rho_{i_{2}, n}^{-1} a_{i_{2} j_{2}} y_{j_{2}, n}\right)\right] \\
& -\left[x_{i_{1}, n} \rho_{i_{1}, n}^{-1} a_{i_{1}, j_{1}} \delta_{j_{1}, n}^{-1} y_{j_{1}, n}+\delta_{j_{2}, n}^{-1}\left(1-x_{i_{1}, n} \rho_{i_{1}, n}^{-1} a_{i_{1}, j_{1}} y_{j_{1}, n}\right)\right] \\
& \leqslant\left[\delta_{j_{2}, n}^{-1} d_{n}+\delta_{j_{1}, n}^{-1}\left(1-d_{n}\right)\right]-\left[\delta_{j_{1}, n}^{-1} d_{n}+\delta_{j_{2}, n}^{-1}\left(1-d_{n}\right)\right] \\
& =\left(1-2 d_{n}\right)\left(\delta_{j_{1}, n}^{-1}-\delta_{j_{2}, n}^{-1}\right) \leqslant\left(1-2 d_{n}\right)\left(\max _{i} \rho_{i, n}-\min _{i} \rho_{i, n}\right) \\
& =\left(1-2 d_{n}\right) \phi\left(x^{n}, y^{n}\right) \leqslant(1-2 \mu) \phi\left(x^{n}, y^{n}\right) .
\end{aligned}
$$

It follows that

$$
\lim _{n \rightarrow \infty} \phi\left(x^{n}, y^{n}\right)=0 .
$$

Since $V$ is bounded, the sequence $\left\{\left(x^{n}, y^{n}\right)\right\}$ has a limit point $(\tilde{x}, \tilde{y})$. It readily follows that $(\tilde{x}, \tilde{y}) \in V$ and $\phi(\tilde{x}, \tilde{y})=0$.

Lemma 2. For any positive matrix $A$, there exist diagonal matrices $D_{1}$ and $D_{2}$ with positive diagonals such that $D_{1} A D_{2}$ and $D_{2} A D_{1}$ are stochastic.

Proof. Pick $(\tilde{x}, \tilde{y}) \in V$, which minimizes $\phi(x, y)$. Then $\sum_{j} \tilde{y}_{i} a_{i j} \tilde{x}_{j}=1$ for all $i$ while $\sum_{j} \tilde{x}_{i} a_{i j} \tilde{y}_{j}=k$, a constant, for all $i$. It readily follows that $k$ and 1 are each maximal eigenvalues for $D_{1} A D_{2}$. Thus $k=1$.

Lemma 3. If $A$ is positive and if $D_{1} A D_{2}$ and $D_{2} A D_{1}$ are both stochastic where $D_{1}$ and $D_{2}$ are diagonal with positive diagonals, then $D_{2}=p D_{1}$ for some number $p>0$.

Proof. Let $D_{1}=\operatorname{dg}\left(x_{1}, \ldots, x_{N}\right)$ and $D_{2}=\operatorname{dg}\left(y_{1}, \ldots, y_{N}\right)$ and set $p_{j}=y_{j} / x_{j}$. If $p_{j}$ is not constant, then

$$
\begin{array}{r}
\max _{j} p_{j}=p_{i 0}=\left(\sum_{j} x_{i_{0}} a_{i_{0} j} x_{j}\right)^{-1}=\left(\sum_{j} x_{i_{0}} a_{i_{0} j} x_{j} p_{j}\right)\left(\sum_{j} x_{i_{0}} a_{i_{0} j} x_{j}\right)^{-1} \\
<\max _{j} p_{j}
\end{array}
$$

a contradiction.
Proof of the theorem. From Lemmas 2 and 3, there is a diagonal matrix $D_{1}$ with positive diagonal and a positive number $p$ such that $p D_{1} A D_{1}$ is stochastic. The existence part of the theorem follows by taking $D=p^{\frac{1}{2}} D_{1}$.

Suppose $D$ and $C$ are diagonal matrices with positive diagonals such that $D A D$ and $C A C$ are stochastic. Then $C A C=\mathrm{B}=\left(b_{i j}\right)$ and $D C^{-1} B C^{-1} D$ are both stochastic. If $D C^{-1}=\operatorname{dg}\left(z_{1}, \ldots, z_{N}\right)$,

$$
\max _{j} z_{j}=z_{i_{0}}=\left(\sum_{j} b_{i_{0} j} z_{j}\right)^{-1} \leqslant\left(\sum_{j} b_{i_{0} j}\right)^{-1}\left(\min _{j} z_{j}\right)^{-1}=\left(\min _{j} z_{j}\right)^{-1}
$$

and similarly $\min _{j} z_{j} \geqslant\left(\max _{j} z_{j}\right)^{-1}$, with equality possible in each case only if $z_{j}$ is constant for all $j$. It follows that $z_{j}=1, j=1, \ldots, N$, and therefore that $D=C$.

Corollary 1 (Marcus and Newman 1). If $A$ is symmetric and has positive entries, there exists a diagonal matrix $D$ with positive main diagonal entries such that DAD is doubly stochastic.

Proof. This follows at once since $D A D$ is symmetric when $A$ is symmetric.
Corollary 2. Corresponding to each positive $N \times N$ matrix $A$ and each set of positive real numbers $p_{1}, \ldots, p_{N}$ there is a unique matrix of the form $D A D$ with row sums $p_{1}, \ldots, p_{N}$ where $D$ is a diagonal matrix with a positive main diagonal.

Proof. Let $D_{0}=\operatorname{dg}\left(p_{1}, \ldots, p_{N}\right)$ and set $B=D_{0}{ }^{-1} A$. There is a diagonal matrix $D$ with a positive main diagonal such that $S=D B D$ is stochastic. Then $D_{0} S=D A D$ has the appropriate row sums. We have used the fact that $D_{0} D=D D_{0}$.

In the proof of Lemma 1 a method is suggested for determining the matrix $D A D$ of the theorem from $A$ by an iterative scheme.

Define diagonal matric sequences $\left\{X_{n}\right\}$ and $\left\{Y_{n}\right\}$ such that

$$
\begin{gathered}
X_{0}=I, \quad Y_{0}=\operatorname{dg}\left[\left(\sum_{j} a_{1 j}\right)^{-1}, \ldots,\left(\sum_{j} a_{N j}\right)^{-1}\right], \\
X_{n+1}=D_{1, n} X_{n}, Y_{n+1}=D_{2, n} Y_{n},
\end{gathered}
$$

where $D_{1, n}$ and $D_{2, n}$ are diagonal matrices such that

$$
D_{1, n}^{-1} u=X_{n} A Y_{n} u \quad \text { and } \quad D_{2, n}^{-1} u=Y_{n} A X_{n+1} u ;
$$

here $u$ denotes the $N$-dimensional vector all of whose components equal one. Then $X_{n} A Y_{n} \rightarrow D A D$.

## References

1. M. Marcus and M. Newman, The permanent of a symmetric matrix, Amer. Math. Soc. Not., 8 (1961), 595.
2. R. Sinkhorn, $A$ relationship between arbitrary positive matrices and doubly stochastic matrices, Ann. Math. Statist., 35 (1964), 876-879.

## University of Houston

