

Appendix C

(The theorem of Perron and Frobenius.)

Irreducible matrix:

A non-negative square matrix is irreducible if its directed graph is strongly connected.

(i.e., if there is a path from every node to every other node).

Graph of a non-negative square matrix $A = (a_{ij})$:

Node j is connected to i if and only if $a_{ij} > 0$.

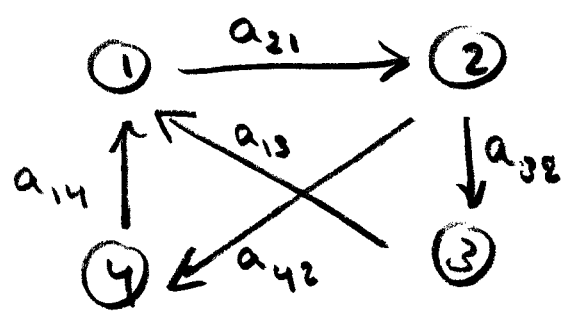
We then write $j \rightarrow i$

Example.

$$A := \begin{pmatrix} 0 & 0 & a_{13} & a_{14} \\ a_{21} & 0 & 0 & 0 \\ 0 & a_{32} & 0 & 0 \\ 0 & a_{42} & 0 & 0 \end{pmatrix}$$

matrix:

$$a_{13}, a_{14}, a_{21}, a_{32}, a_{42} > 0$$



Graph

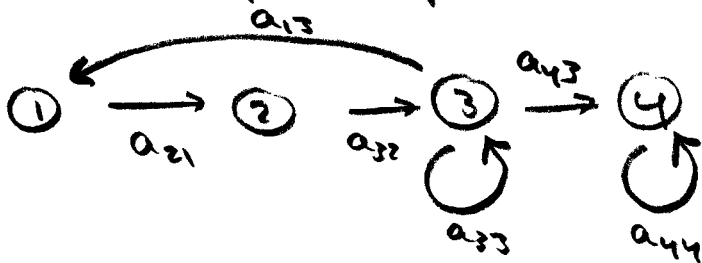
The graph is obviously strongly connected.

So, the matrix is irreducible.

Example

Age-structured population
with part-reproductive
individuals (state 4).

life-cycle graph:



Not strongly connected, because
from 4 you cannot go to 1, 2
or 3.

Matrix:

$$A = \begin{pmatrix} 0 & 0 & a_{13} & 0 \\ a_{21} & 0 & 0 & 0 \\ 0 & a_{32} & a_{33} & 0 \\ 0 & 0 & a_{43} & a_{44} \end{pmatrix}$$

So, the corresponding matrix
is not irreducible (i.e., reducible)

Theorem

$A \in \mathbb{R}^{d \times d}$ is irreducible if and only if $(I+A)^{d-1}$ is strictly positive.

Proof:

(See: Horn & Johnson, 1985, pp. 507-520.)

Primitive matrix

A non-negative matrix A is primitive if it becomes strictly positive if raised to a sufficiently high power.

(i.e., if $\exists k > 0 : A^k > 0$).

Properties

- Any primitive matrix is irreducible.
- An irreducible matrix is primitive if the greatest common division of the lengths of all loops in the directed graph is equal to one,
(\rightarrow Rosenblatt, 1957).

Theorem

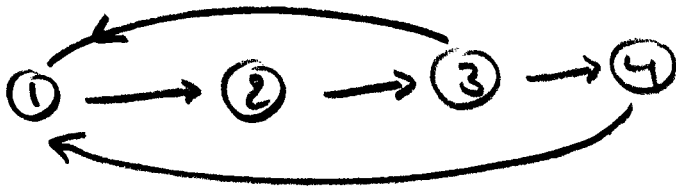
$A \in \mathbb{R}^{d \times d}$ is primitive if and only if $A^{d^2 - 3d + 2}$ is strictly positive.

Proof

See: Horn & Johnson, 1985,

pp. 507-520.

Example.



graph is strongly connected and

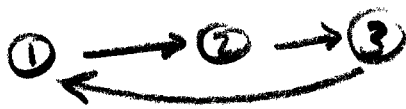
there are at least two loops

(namely $1 \rightarrow 2 \rightarrow 3$ and $1 \rightarrow 2 \rightarrow 3 \rightarrow 4$)

of lengths 3 and 4, which have a common divisor equal to one.

⇒ The corresponding matrix is primitive.

Example



Strongly connected, but the greatest common divisor is 3.

⇒ Corresponding matrix is reducible but not primitive.

Perron-Frobenius (P.F.)

This is part of the P.F. theorem
Proof in, e.g., Horn & Johnson
1985

① Suppose $A \in \mathbb{R}^{d \times d}$ is non-negative
and primitive.

Then there exists an eigenvalue $\lambda_1 > 0$ which is a simple root of the characteristic equation, and which has associated left and right eigenvectors $v_1 > 0$ and $w_1 > 0$ and all other eigenvalues λ_i ($i \geq 2$) satisfy $\lambda_1 > |\lambda_i|$.

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of the characteristic equation and which has associated left and right eigenvectors $v_1 > 0$ and $w_1 > 0$, and all other eigenvalues λ_i ($i \geq 2$) satisfy

$\lambda_1 \geq |\lambda_i|$, but if $\lambda_1 = |\lambda_i|$ for some $i \neq 1$, then λ_i is complex and has complex eigenvectors.

Dominant eigenvalue

The eigenvalue λ_1 above is called the dominant eigenvalue.

Application (discrete time)

Consider the discrete time invader dynamics.

$$\boxed{u(t+1) = A u(t)} \quad u \in \mathbb{R}^d \quad (d > 1)$$

where A is a constant, non-negative and primitive matrix, with eigenvalues $\lambda_1, \dots, \lambda_d$ and corresponding right eigenvectors w_1, \dots, w_d .

From the PF theorem we know that A has a dominant eigenvalue, which we can choose to be denoted λ_1 .

Write

$$\Lambda = \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_d \end{pmatrix} \quad \text{and} \quad W = (w_1, \dots, w_d)$$

Both are $d \times d$ matrices, and

$$\boxed{AW = W\Lambda}$$

Assume that W is nonsingular (i.e., w_1, \dots, w_d are linearly independent)

Then

$$A = W \Lambda W^{-1}$$

and hence

$$W^{-1} m(t+1) = \Lambda \cdot W^{-1} m(t)$$

which is a system of decoupled difference equations with solution

$$W^{-1} m(t) = \Lambda^t W^{-1} m(0)$$

and so

$$m(t) = W \Lambda^t W^{-1} m(0)$$

(*)

where

$$\Lambda^t = \begin{pmatrix} \lambda_1^t & 0 \\ 0 & \lambda_d^t \end{pmatrix}$$

Since the eigenvectors w_1, \dots, w_d are assumed to be lin. indep. there exists a (complex-valued) vector $c = (c_1, \dots, c_d)$ such that

$$m(0) = c_1 w_1 + \dots + c_d w_d = Wc$$

Substitution into (*) gives

$$u(t) = W \Lambda^t W^{-1} Wc = W \Lambda^t c$$

that is,

$$u(t) = \sum_{i=1}^n c_i \lambda_i^t w_i$$

Since $\lambda_1 > 0$ (P.F.) we can write

$$\frac{u(t)}{\lambda_1^t} = c_1 w_1 + \sum_{i \geq 2} c_i \left(\frac{\lambda_i}{\lambda_1}\right)^t w_i$$

and since $\lambda_1 > |\lambda_i| \forall i \geq 2$ (P.F.) $(\lambda_i / \lambda_1)^t \rightarrow 0$ as $t \rightarrow \infty$, and hence

(**) $\lim_{t \rightarrow \infty} \frac{u(t)}{\lambda_1^t} \rightarrow c_1 w_1$

In other words, the structure of the population converges to the dominant eigenvector as $t \rightarrow \infty$, which is known as the strong ergodic theorem

Taking norms and logarithms in (**) and then dividing by t gives

$$\left(\lim_{t \rightarrow \infty} \frac{\log N_{i,t}}{t} - \log \lambda_1 = 0 \right)$$

invasion
fitness

⇒

invasion fitness	=	<u>logarithm</u> dominant eigenvalue
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(discrete time)

Application (cont. time)

$$\dot{m} = A m \quad | \quad m \in \mathbb{R}^d \quad (d \geq 2)$$

where A is a constant matrix with non-negative off-diagonal elements, and eigen values $\lambda_1, \dots, \lambda_d$ and corresponding right-eigen vectors w_1, \dots, w_d .

Let $\mu > 0$ such that $A + \mu I$ is non-negative (i.e., also on the diagonal) and suppose $A + \mu I$ is irreducible.

↑
note, that this is a condition on A (not on μ), and therefore we can also say that A is irreducible even if its diagonal elements may be negative

The eigenvalues of $A + \mu I$ are $\lambda_1 + \mu, \dots, \lambda_d + \mu$ and the associated eigenvectors w_1, \dots, w_d .

By the P.F. theorem we know that $A + \mu I$ has a dominant eigenvalue, which we can take to be $\lambda_1 + \mu$. Hence, for all $i \neq 1$:

$$\begin{aligned} |\lambda_1| &= (\lambda_1 + \mu) - \mu \stackrel{\text{P.F. (2)}}{\geq} |\lambda_i + \mu| - \mu \stackrel{\text{triangular inequality}}{\geq} \\ &\geq |\operatorname{Re}(\lambda_i + \mu)| - \mu \stackrel{\text{P.F. (2)}}{>} \operatorname{Re}(\lambda_i + \mu) - \mu = \operatorname{Re} \lambda_i \end{aligned}$$

To show:

$$(*) \quad \boxed{\lambda_1 > \operatorname{Re} \lambda_i \quad \forall i \neq 1}$$

Solving $u' = Au$ gives

$$(**) \quad \boxed{u(t) = We^{\Lambda t} W^{-1} u(0)}$$

where

$$\Lambda = \begin{pmatrix} \lambda_1 & & 0 \\ & \ddots & \\ 0 & & \lambda_d \end{pmatrix} \quad \text{and} \quad W = (w_1, \dots, w_d)$$

and assuming that W is not singular.

Write $u(t)$ as a linear combination of w_1, \dots, w_d :

$$u(t) = \sum_{i=1}^d w_i \cdot c_i = Wc$$

where $c = (c_1, \dots, c_d) \in \mathbb{C}^d$.

Then, from $(*)$ on prev. page,

$$u(t) = W e^{\Lambda t} c = \sum_{i=1}^d c_i e^{\lambda_i t} w_i$$

$\text{Re}(\lambda_i - \lambda_1) < 0$

Hence

$$e^{-\lambda_1 t} u(t) = c_1 w_1 + \sum_{i \geq 2} c_i e^{(\lambda_i - \lambda_1)t} w_i$$

$\rightarrow c_1 w_1$ as $t \rightarrow \infty$.

So, the structure of the invader population converges to the dominant eigenvector as $t \rightarrow \infty$

(Strong ergodic theorem).

$$e^{-\lambda_1 t} \|u(t)\| \rightarrow c \|w\| \quad \text{as } t \rightarrow \infty$$

Taking norms, logarithms and dividing by t gives

$$\underbrace{\frac{\log \|u(t)\|}{t}}_{\text{invasion fitness}} - \lambda_1 \rightarrow 0 \quad \text{as } t \rightarrow \infty$$

In other words:

$$\text{invasion fitness} = \text{dominant eigenvalue}$$

(continuous time)