# 'MATH+ECON+CODE' MASTERCLASS ON EQUILIBRIUM TRANSPORT AND MATCHING MODELS IN ECONOMICS

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Special lecture 2. Perron-Froebenius theory

#### REFERENCES

- Bertsekas, Tsitsiklis. Parallel and distributed computation: Numerical methods.
- ▶ Berman, Plemmons. Nonnegative matrices in the mathematical sciences.
- ► Tsatsomeros. Lecture Notes on Matrices with Positive Principal Minors: Theory and Applications. (Online)

## **JACOBI FOR MATRICES**

► Consider the equilibrium problem

$$Qp = q$$

where Q is a  $n \times n$  matrix. Assume  $Q_{ii} > 0$  for each i. (In fact this will be implied by stronger assumptions).

- ▶ When Q is invertible (more on this later), and denote  $p^*$  the solution of the above equation.
- ▶ We shall discuss methods to look for  $p^*$ , in the presence of gross substitutes.

# GROSS SUBSTITUTES AND Z-MATRICES

- $\triangleright$   $(Qp)_i$  is interpreted as the supply for good i.  $Q_{ii} > 0$  means that when the price of good i increases, the supply for it increases.
- Assume gross substitutes, that is  $Q_{ij} \leq 0$  for  $i \neq j$ . Interpretation: when the price of good j increases, the production of good i decreases because suppliers substitute producing j to producing i.

**Definition**. One says Q is a Z-matrix when  $Q_{ij} \leq 0$  for  $i \neq j$ .

► In the literature, Z-matrices are sometimes referred to as negative *Metzler matrices*. (Metzler matrices are non-negative off-diagonal).

# JACOBI ALGORITHM

▶ Recall what the Jacobi algorithm is. Decompose *Q* as

$$Q = \Delta - A$$

where  $\Delta$  is diagonal with positive entries, and A has nonnegative terms and zeros on the diagonal.

► Jacobi algorithm rewrites as

$$\Delta p^{k+1} - Ap^k = q$$

that is

$$p^{k+1} = \Delta^{-1}Ap^k + \Delta q.$$

▶ As a result, when  $p^*$  exists, setting  $\delta^k = p^k - p^*$ , we have

$$\delta^k = \left(\Delta^{-1}A\right)^k \delta^0,$$

and we wonder when Jacobi converges for any starting point  $p^0$ .

## THE ROLE OF POSITIVE EIGENVECTORS

Consider v an eigenvector of  $M = \Delta^{-1}A$  and assume  $v_i > 0$  for all i. Then the associated eigenvalue  $\lambda$  is > 0. We have for any  $\delta$ 

$$(M\delta)_{i} = \sum_{j} M_{ij} \delta_{j} = \sum_{j} M_{ij} v_{j} \frac{\delta_{j}}{v_{j}} \leq (Mv)_{i} \max_{j} \left( \left| \frac{\delta_{j}}{v_{j}} \right| \right) = \lambda v_{i} \left| \delta \right|_{v}^{\infty}$$

where

$$|\delta|_{v}^{\infty} := \max_{j} \left( \left| \frac{\delta_{j}}{v_{j}} \right| \right)$$

► As a result,

$$|M\delta|_{v}^{\infty} \leq \lambda |\delta|_{v}^{\infty}$$

and thus, if  $\lambda < 1$ ,  $M^k \delta \to 0$  for any  $\delta$ .

#### SPECTRAL RADIUS AND INDUCED NORMS

▶ Given a norm |x|, the *induced norm* ||.|| on matrices is defined as

$$||M|| = \max\{|Mx| : |x| = 1\}$$

- ▶ The spectral radius  $\rho$  (M) as the maximum modulus of the (complex) eigenvalues of M.
- ► While the induced norm depends on the norm that is chosen, the spectral radius does not. We have easily

$$\rho\left(M\right) \leq \|M\|$$

for any induced norm, and (less easily) Gelfand's formula

$$\rho\left(M\right) = \lim_{k \to \infty} \left\| M^k \right\|^{1/k}$$

▶ When *M* is symmetric, the spectral radius coincides with the induced Euclidean norm, which is itself an Euclidean norm on matrices. Thus, the following developments have interest only outside of that case.

#### SPECTRAL RADIUS

- ▶ A matrix M is convergent if  $M^k \to 0$  as  $k \to +\infty$ . We have: **Proposition** (BT prop. A.20): M is convergent if and only if  $\rho(M) < 1$ .
- ▶ As a result, if  $\|M\| < 1$  for some induced norm, then M is convergent, but the converse is not true.
- ▶ However, we shall see that when the Perron-Froebenius theorem applies on M, then there exists a norm |.| such that  $\rho\left(M\right) = \|M\|$  for  $\|.\|$  the induced matrix norm.

#### IRREDUCTIBLE MATRICES

Before that, we need an important definition.

**Definition**. A matrix M is irreductible iff for every i and j there is a path  $i_0 = i, ... i_p = j$  such that  $M_{i_k i_{k+1}} \neq 0$ .

Note that in our example with  $Q=\Delta-A$ , Q has connected strong substitutes if and only if  $M=\Delta^{-1}A$  is irreductible.

# THE PERRON-FROEBENIUS THEOREM

We have seen that Jacobi converges if and only if  $ho\left(\Delta^{-1}A\right)<1.$ 

 $\Delta^{-1}A$  being a matrix with nonnegative components, we need a result on spectrum of nonnegative matrices. The Perron-Froebenius applies to that. **Theorem (BT Prop. 6.6)**. Let M be a  $n \times n$  matrix with nonnegative

terms with is irreductible. Then:

- ho (M) is an eigenvalue of M, and there exists a associated right eigenvector v with positive entries (that is, there exists v such that  $Mv = \rho(M) v$  and  $v_i > 0$  for all i).
- ▶ *v* above is (up to rescaling) the only eigenvector of *M* with positive entries. It is the so-called left Perron eigenvector.
- ▶ The rank of  $M \rho(M)I$  is n 1.
- ▶ Furthermore, considering  $|z|_v^\infty = \max\{|z_i/v_i|\}$ , and denoting  $\|M\|_v^\infty$  the matrix norm induced by that norm, one has

$$\rho\left(M\right)=\left\|M\right\|_{v}^{\infty}.$$

## AN ASIDE: MARKOV CHAINS

- ▶ Let M be a  $n \times n$  matrix. This is viewed as the matrix of Markov transitions of a Markov chain on state space= $\{1, ..., n\}$ , where  $M_{ij}$  is the probability of visiting i at next step conditional on being at j at the current step. We impose therefore that  $M_{ii} \geq 0$ , and  $\sum_i M_{ii} = 1$ .
- ▶ M is irreductible means that for any  $i \neq j$  there is a k such that  $(M^k)_{ij} > 0$  meaning that if you wait long enough, you have a positive probability of visiting every state conditional on being in any state.

## MARKOV CHAINS AND STATIONARY DISTRIBUTIONS

- ▶ Because  $M^{\top}1_n = 1_n$ , 1 is an eigenvalue of  $M^{\top}$  with associated eigenvector  $1_n$ . By Perron-Froebenius, this implies that 1 is the largest eigenvalue of  $M^{\top}$  hence of M, that is  $\rho(M) = 1$ .
- **b** By Perron-Froebenius again, M has an eigenvector with positive components, call it  $\pi_i > 0$  associated with eigenvalue 1. This means

$$M\pi = \pi$$

and we can impose  $\sum_i \pi_i = 1$ .

► That is

$$\sum_{j} M_{ij} \pi_{j} = \pi_{j}$$

hence  $\pi$  can be interpreted as the stationary distribution of the Markov chain.

## NONREVERSING MATRICES

**Definition**. A matrix M is *nonreversing* if  $\delta \geq 0$  and  $M\delta \leq 0$  imply that  $\delta = 0$ .

# Remarks:

► "Nonreversing" is not a standard terminology. In Tsatsomeros' terminology, it is equivalent with " $-M^{\top}$  is not semipositive".

#### M-MATRICES

**Definition**. A M-matrix is a Z-matrix which is nonreversing.

The Twenty Equivalence theorem. (BP theorem 4.6). Assume M is a Z-matrix. Then the following statements are equivalent to "M is a M-matrix":

- (1)  $M^{-1}$  is entrywise positive
- (2) Jacobi converges from any starting point
- (3)  $\rho \left( \Delta^{-1} A \right) < 1$ .
- (4) There exists a vector  $w_i > 0$  such that diag(w) M is diagonally dominating.
- plus over 17 equivalences...

## LAW OF AGGREGATE SUPPLY

- As  $\rho(M) = \rho(M^{\top})$ , the result of the Perron-Froebenius theorem can be applied to  $M^{\top}$ , and there is a left eigenvector u with positive entries such that  $M^{\top}u = \rho(M)u$ .
- ► What does this entails economically?
- ► Consider  $M = \Delta^{-1}A$  and set  $\lambda = \rho (\Delta^{-1}A)$ . We have  $A^{\top}\Delta^{-1}u = \lambda u$ , and therefore, setting  $w = \Delta^{-1}u$ , we have

$$A^{\top}w = \lambda \Delta w$$

and hence

$$w^{\top}Q = w^{\top}(\Delta - A) = (1 - \lambda) w^{\top}\Delta$$

▶ This means that

$$\sum_{i} w_{i} Q_{ij} = (1 - \lambda) w_{j} Q_{jj}$$

which implies that the matrix diag(w)Q is diagonally dominating.

► This implies the weighted law of aggregate supply:

$$\sum_{i} w_{i} (Qp)_{i} = (1 - \lambda) \sum_{i} w_{i} Q_{ii} p_{i}$$

is a increasing function in each of the  $p_i$ .