

# ‘math+econ+code’ masterclass on equilibrium transport and matching models in economics

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Special lecture 1. Perron-Frobenius theory

- ▶ 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)

- 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.

- ▶  $(Qp)_i$  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).

- 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  $p^0$ .

- 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  $\delta$

$$(M\delta)_i = \sum_j M_{ij}\delta_j = \sum_j M_{ij}v_j \frac{\delta_j}{v_j} \leq (Mv)_i \max_j \left( \left| \frac{\delta_j}{v_j} \right| \right) = \lambda v_i |\delta|_v^\infty$$

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  $\lambda < 1$ ,  $M^k \delta \rightarrow 0$  for any  $\delta$ .

- ▶ Given a norm  $|x|$ , the *induced norm*  $\|\cdot\|$  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

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

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

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

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

- ▶ A matrix  $M$  is *convergent* if  $M^k \rightarrow 0$  as  $k \rightarrow +\infty$ . 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  $|\cdot|$  such that  $\rho(M) = \|M\|$  for  $\|\cdot\|$  the induced matrix norm.



Before that, we need an important definition.

**Definition.** A matrix  $M$  is irreducible iff for every  $i$  and  $j$  there is a path  $i_0 = i, \dots, i_p = j$  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 irreducible.

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

$\Delta^{-1}A$  being a matrix with nonnegative components, we need a result on spectrum of nonnegative matrices. The Perron-Frobenius applies to that.

**Theorem (BT Prop. 6.6).** Let  $M$  be a  $n \times n$  matrix with nonnegative terms with is irreducible. Then:

- ▶  $\rho(M)$  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

$$\rho(M) = \|M\|_v^\infty.$$

- ▶ Let  $M$  be a  $n \times n$  matrix. This is viewed as the matrix of Markov transitions of a Markov chain on state space  $= \{1, \dots, n\}$ , 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  $M_{ij} \geq 0$ , and  $\sum_i M_{ij} = 1$ .
- ▶  $M$  is irreducible 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.

- ▶ Because  $M^\top \mathbf{1}_n = \mathbf{1}_n$ , 1 is an eigenvalue of  $M^\top$  with associated eigenvector  $\mathbf{1}_n$ . By Perron-Frobenius, this implies that 1 is the largest eigenvalue of  $M^\top$  hence of  $M$ , that is  $\rho(M) = 1$ .
- ▶ By Perron-Frobenius 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

$$\sum_j M_{ij} \pi_j = \pi_i$$

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

**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”.

**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)  $\rho(\Delta^{-1}A) < 1$ .
  - (4) There exists a vector  $w_i > 0$  such that  $\text{diag}(w) M$  is diagonally dominating.
- plus over 17 equivalences...

- ▶ As  $\rho(M) = \rho(M^\top)$ , the result of the Perron-Frobenius 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  $\text{diag}(w) Q$  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  $p_i$ .