# SPECTRAL THEORY OF REDUCIBLE NONNEGATIVE MATRICES IN MAX ALGEBRA

Hans Schneider

Chemnitz October 2010

maxspectchmn 21 Sept 2010, 14:30

## max, min, +, times

```
\begin{array}{c} \text{max plus } (\mathbb{R}, \text{max}, +) \\ \text{max times } (\mathbb{R}_0^+, \text{max}, \times) \\ \text{min plus } (\mathbb{R}, \text{min}, +) = \text{tropical} \\ \text{min times } (\mathbb{R}_0^+, \text{min}, \times) \end{array}
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We do max times

#### **MAX ALGEBRA**

$$a, b \ge 0$$
  
 $a \oplus b = \max(a, b)$   
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 $(\mathbb{R}^+_0,\oplus,\otimes)$  is a semiring: a commutative semigroup with 0 under max, and  $\otimes$  distributes over  $\oplus$  Just like  $(\mathbb{R}^+_0,+,\times)$ ?

$$a \oplus b = 0 \Longrightarrow a = b = 0$$

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The spectral theory in **MX** "is extremely similar to the well-known Perron-Frobenius theory" in **NN** with some important differences.

Our aim is to compare and contrast the two theories

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$$A \in \mathbb{R}_+^{n \times n}, \ A \ge 0$$

G(A): Graph of AVertex set  $\{1, ..., n\}$ arcs  $i \rightarrow j$  :  $a_{ij} > 0$ 

$$A \in \mathbb{R}^{n \times n}_+, \ A \ge 0$$

$$\mathcal{G}(A) \colon \text{ Graph of } A$$

$$\text{ Vertex set } \{1, \dots, n\}$$

$$\text{ arcs } i \to j \quad : \quad a_{ij} > 0$$

$$i_0 \stackrel{*}{\to} i_k : \quad \exists (i_1, \dots, i_{k-1}) \quad i_0 \to i_1 \dots \to i_{k-1} \to i_k$$

$$\text{ or } \quad i = j$$

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$$\text{ cycle mean } \bar{\gamma}(A) = (a_{i_0, i_1} \dots a_{i_{k-1}, i_k})^{1/k}$$

$$\rho(A) = \max \bar{\gamma}(A), \gamma(A) \in cG(A)$$

$$A = \begin{pmatrix} 3/4 & 1 & 1/2 & 0 & 0 & 1/2 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 3/4 & 0 & 0 \\ 1/2 & 0 & 0 & 0 & 1 & 3/4 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1/2 \\ 3/4 & 0 & 0 & 0 & 0 & 3/4 & 0 \end{pmatrix}$$

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$$\rho(A) = 1$$
Components of  $C(A)$ :  $\{1, 2, 3, 4\}, \{5\}$ 

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Very special?

Very special? Not really! Very special? Not really!

Observation: If  $B = X^{-1}AX$ , where X is a pos diag matrix then

$$b_{ij} = rac{x_i a_{ij}}{x_j}$$
 $\mathcal{G}(B) = cG(A)$ 
 $ar{\gamma}(A) = ar{\gamma}(B), \ orall \ ext{cycles } \gamma$ 
 $ho(B) = 
ho(A)$ 
 $\mathcal{C}(B) = cC(A)$ 

# diagonal scaling

Fiedler-Ptak(1967, 1969), M.Schneider -S (1990)

## Theorem

Let  $A \in \mathbb{R}^{n \times n}_+$ . There exists a pos diag X such that for  $B = X^{-1}AX$ ,

$$b_{ij} = \rho(B)$$
 if  $(i,j) \in C(B)$   
 $b_{ii} < \rho(B)$  otherwise

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We may assume matrix is strictly visualized

#### NN: Perron-Frobenius for irred matrices

Frobenius (1912)

## Theorem

Let  $A \ge 0$  be irreducible. Then its spec rad  $\rho(A)$  is its the unique eigenvalue with an assoc nonneg eigenvector

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 $\rho(A)$  is the **Perron root** of A

#### **MX: Perron-Frobenius for irred matrices**

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#### **MX: Perron-Frobenius for irred matrices**

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## Theorem

Let  $A \ge 0$  be irreducible. Then its max cyc mean  $\rho(A)$  is its unique (dist) eigenvalue. There is an (ess) unique associated positive eigenvector for each component of the crit graph, which are the extremals of the max cone of eigenvectors.

 $\rho(A)$  will be called the **(max) Perron root** of A

# MX: Example

$$A = \begin{pmatrix} 3/4 & 1 & 1/2 & 0 & 0 & 1/2 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 3/4 & 0 & 0 \\ 1/2 & 0 & 0 & 0 & 1 & 3/4 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1/2 \\ 3/4 & 0 & 0 & 0 & 0 & 3/4 & 0 \end{pmatrix} \quad \begin{array}{c} 1 & 3/4 \\ 1 & 3/4 \\ 1 & 3/4 \\ 1/2 & 1 \\ 3/8 & 9/32 \\ 3/4 & 9/16 \\ \end{array}$$

Two evectors of  $\rho(A) = 1$ 

#### **Frobenius Normal Form**

collect strong conn cpts [classes] of  $\mathcal{G}(A)$  and linearly order them

After permutation similarity

$$A = \begin{bmatrix} A_{11} & 0 & \dots & \dots & 0 \\ A_{21} & A_{22} & 0 & \dots & 0 \\ \vdots & \vdots & \ddots & \ddots & \vdots \\ \vdots & \vdots & & \ddots & 0 \\ A_{k1} & A_{k2} & \dots & \dots & A_{kk} \end{bmatrix}$$

each diagonal block irreducible

## reduced graph

Reduced graph  $\mathcal{R}(A)$ 

$$V = \{1, \ldots, k\}$$

$$i \rightarrow j \in E: A_{ij} \neq 0$$

Path from *i* to *j* 

$$i_0 \rightarrow i_1 \rightarrow \cdots \rightarrow i_{p-1} \rightarrow i_m$$

Transitive closure  $\mathcal{R}^*(A)$ 

$$i \stackrel{*}{\rightarrow} j$$
: exists path from *i* to *j*

Skeleton  $S = \mathcal{R}_*(A)$ 

$$(i,j) \in S: i \xrightarrow{*} k \xrightarrow{*} j \text{ implies } k = i \text{ or } k = j$$

### **Example**

$$\begin{array}{cccc} (1) & \longleftarrow & (2) & \longleftarrow & (3) \\ ? \nwarrow & \uparrow & \\ & & (4) & \end{array}$$

## **Example**

- irred block
- nonzero block
- in trans closure of skeleton

## Marked Reduced graph $\mathcal{R}(A)$

Vertex set 
$$\{1, ..., k\}$$
 (classes)

$$i \rightarrow j \iff A_{ij} \geq 0$$

*j* has access to *i* in  $\mathcal{R}(A)$ :

$$i \stackrel{*}{\leftarrow} j$$

Each vertex marked with its (max) Perron root

i distinguished

$$i \stackrel{*}{\leftarrow} j \Longrightarrow \rho_i > \rho_j$$

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2,3,4 distinguished, 1 semi-distinguished

## MX: eigenvalues, eigenvectors

Gaubert (1990s), Butkovic&Cuninghame-Green&Gaubert(2009)

#### **Theorem**

Let A be a nonnegative matrix in FNF. Then  $\lambda$  is an evalue of (A) if snd only there is a semi-distinguished vertex i with  $\rho_i = \lambda$ 

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The eigenvectors of A correspond to the semi-distinguished vertices of A: for for each semi-distinguished vertex i of  $\mathcal{R}(A)$  there are (nonnegative) eigenvectors  $x^i$  with  $Ax^i = \rho_i x^i$  such that

$$x_j^i > 0$$
 if  $i \leftarrow \leftarrow j$   
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 $x'_j = 0$  otherwise

Properly chosen, these form the exremals oif the cones of eignevectors

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#### **Theorem**

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 if  $i \leftarrow \leftarrow j$   
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Properly chosen, these are linearly independent, and for any part evalue, form the extremals of the cone of nonneg evectors

## NN example

$$A = \begin{pmatrix} 4 & 0 & 0 & 0 \\ 1 & 4 & 1 & 0 \\ 0 & 3 & 2 & 0 \\ 0 & 0 & 3 & 2 \end{pmatrix}$$

$$[4] \leftarrow [5]^{**} \leftarrow [2]^{**}$$

$$\begin{array}{cccc} 0 & . & 0 \\ 1 & . & 0 \\ 1 & . & 0 \\ 1 & . & 1 \end{array}$$

## MX example

$$A = \begin{pmatrix} 4 & 0 & 0 & 0 \\ 1 & 4 & 1 & 0 \\ 0 & 3 & 2 & 0 \\ 0 & 0 & 3 & 2 \end{pmatrix}$$
$$\begin{pmatrix} [4]^* & \leftarrow & [4]^{**} & \leftarrow [2]^{**} \end{pmatrix}$$
$$\begin{pmatrix} 1 & 0 & 0 \\ 1/4 & 1 & 0 \\ 3/16 & 3/4 & 0 \\ 9/64 & 3/8 & 1 \end{pmatrix}$$

## P. Butkovic Max-Linear Systems: Theory and Algorithms Springer 2010

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That's it for today

next time:

Commuting matrices in three incarnations

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THANKS!