Tung T. Nguyen

October 17th, 2021

First SIBAU-NU Workshop on Matrix Analysis and Linear Algebra Joint work with Jackie Doan, Jan Minac, Lyle Muller, Federico Pasini

## Social network



Photo credit: Analytics Vidhya

A directed graph is an ordered pair G = (V, E) where

• *V* is a finite set whose elements are called vertices.

A directed graph is an ordered pair G = (V, E) where

- *V* is a finite set whose elements are called vertices.
- *E* is a set of directed edges between vertices.

A directed graph is an ordered pair G = (V, E) where

- V is a finite set whose elements are called vertices.
- E is a set of directed edges between vertices.

Suppose the vertex set of G is  $\{v_1, v_2, \dots, v_n\}$ . A convenient way to represent G is to use its adjacency matrix  $A = (A_{ij})$  where

$$A_{ij} = \begin{cases} 1 & \text{if } v_i \to v_j \\ 0 & \text{else.} \end{cases}$$

A directed graph is an ordered pair G = (V, E) where

- V is a finite set whose elements are called vertices.
- *E* is a set of directed edges between vertices.

Suppose the vertex set of G is  $\{v_1, v_2, \dots, v_n\}$ . A convenient way to represent G is to use its adjacency matrix  $A = (A_{ij})$  where

$$A_{ij} = \begin{cases} 1 & \text{if } v_i \to v_j \\ 0 & \text{else.} \end{cases}$$

The spectrum of G, denoted by Spec(G), is the set of all eigenvalues of its adjacency matrix A.

A directed graph is an ordered pair G = (V, E) where

- V is a finite set whose elements are called vertices.
- E is a set of directed edges between vertices.

Suppose the vertex set of G is  $\{v_1, v_2, \dots, v_n\}$ . A convenient way to represent G is to use its adjacency matrix  $A = (A_{ij})$  where

$$A_{ij} = egin{cases} 1 & ext{if } v_i 
ightarrow v_j \ 0 & ext{else}. \end{cases}$$

The spectrum of G, denoted by Spec(G), is the set of all eigenvalues of its adjacency matrix A. Equivalently, it is the set of all roots of the characteristic polynomial  $p_A(t)$  of A where

$$p_A(t) = \det(tI_n - A).$$

Graphs play a fundamental roles in many real-life applications. Some of them are

 Spectral graph theory is important for many search algorithms such as the Google's famous PageRank algorithm.

- Spectral graph theory is important for many search algorithms such as the Google's famous PageRank algorithm.
- Spectral graph theory has been widely used to analyze diverse data sets in the fields of computational chemistry and bioinformatics.

- Spectral graph theory is important for many search algorithms such as the Google's famous PageRank algorithm.
- Spectral graph theory has been widely used to analyze diverse data sets in the fields of computational chemistry and bioinformatics.
- Many spectral clustering techniques use the spectrum to perform dimensionality reduction on big data sets.

- Spectral graph theory is important for many search algorithms such as the Google's famous PageRank algorithm.
- Spectral graph theory has been widely used to analyze diverse data sets in the fields of computational chemistry and bioinformatics.
- Many spectral clustering techniques use the spectrum to perform dimensionality reduction on big data sets.
- In our study of networks of oscillators, the topology of the networks (i.e. the structure of connections) plays a fundamental role in understanding the existence and stability of the oscillators.

- Spectral graph theory is important for many search algorithms such as the Google's famous PageRank algorithm.
- Spectral graph theory has been widely used to analyze diverse data sets in the fields of computational chemistry and bioinformatics.
- Many spectral clustering techniques use the spectrum to perform dimensionality reduction on big data sets.
- In our study of networks of oscillators, the topology of the networks (i.e. the structure of connections) plays a fundamental role in understanding the existence and stability of the oscillators.
- And much more!

A special type of network that appears often in our work in computational neuroscience is the ring graph.

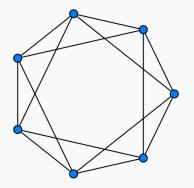


Photo credit: Wikimedia

The adjacency matrix of this graph is given by

```
\begin{bmatrix} 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 1 & 0 & 0 & 1 \\ 1 & 1 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 1 & 1 \\ 1 & 0 & 0 & 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & 1 & 0 \end{bmatrix}.
```

The adjacency matrix of this graph is given by

$$\begin{bmatrix} 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 1 & 0 & 0 & 1 \\ 1 & 1 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 1 & 1 \\ 1 & 0 & 0 & 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & 1 & 0 \end{bmatrix}.$$

We see that all row vectors are composed of the same elements and each row vector is rotated one element to the right relative to the preceding row vector.

The adjacency matrix of this graph is given by

$$\begin{bmatrix} 0 & 1 & 1 & 0 & 0 & 1 & 1 \\ 1 & 0 & 1 & 1 & 0 & 0 & 1 \\ 1 & 1 & 0 & 1 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 1 & 0 \\ 0 & 0 & 1 & 1 & 0 & 1 & 1 \\ 1 & 0 & 0 & 1 & 1 & 0 & 1 \\ 1 & 1 & 0 & 0 & 1 & 1 & 0 \end{bmatrix}.$$

We see that all row vectors are composed of the same elements and each row vector is rotated one element to the right relative to the preceding row vector. This is an example of a circulant matrix.

More generally, a circulant matrix is a matrix of the form

$$C = \begin{bmatrix} c_0 & c_1 & \cdots & c_{n-2} & c_{n-1} \\ c_{n-1} & c_0 & c_1 & & c_{n-2} \\ \vdots & c_{n-1} & c_0 & \ddots & \vdots \\ c_2 & & \ddots & \ddots & c_1 \\ c_1 & c_2 & \cdots & c_{n-1} & c_0 \end{bmatrix}.$$

In particular, a circulant matrix is completely determined by the first row vector  $\vec{c} = (c_0, c_1, \dots, c_{n-1})$ .

# The Circulant Diagonalization Theorem

Let us take a concrete example of a circulant matrix of size  $3 \times 3$ .

$$C = \begin{pmatrix} c_0 & c_1 & c_2 \\ c_2 & c_0 & c_1 \\ c_1 & c_2 & c_0 \end{pmatrix}.$$

# The Circulant Diagonalization Theorem

Let us take a concrete example of a circulant matrix of size  $3 \times 3$ .

$$C = egin{pmatrix} c_0 & c_1 & c_2 \ c_2 & c_0 & c_1 \ c_1 & c_2 & c_0 \end{pmatrix}.$$

Let  $\omega_3$  be 3-root of unity. Then we have

$$C\begin{pmatrix} 1\\ \omega_3\\ \omega_3^2 \end{pmatrix} = \begin{pmatrix} c_0 + c_1\omega_3 + c_2\omega_3^2\\ c_2 + c_0\omega_3 + c_1\omega_3^2\\ c_1 + c_2\omega_3 + c_0\omega_3^2 \end{pmatrix} = \begin{pmatrix} (c_0 + c_1\omega_3 + c_2\omega_3^2)1\\ (c_0 + c_1\omega_3 + c_2\omega_3^2)\omega_3\\ (c_0 + c_1\omega_3 + c_2\omega_3^2)\omega_3^2 \end{pmatrix}.$$

# The Circulant Diagonalization Theorem

Let us take a concrete example of a circulant matrix of size  $3 \times 3$ .

$$C = \begin{pmatrix} c_0 & c_1 & c_2 \\ c_2 & c_0 & c_1 \\ c_1 & c_2 & c_0 \end{pmatrix}.$$

Let  $\omega_3$  be 3-root of unity. Then we have

$$C\begin{pmatrix} 1\\ \omega_3\\ \omega_3^2 \end{pmatrix} = \begin{pmatrix} c_0 + c_1\omega_3 + c_2\omega_3^2\\ c_2 + c_0\omega_3 + c_1\omega_3^2\\ c_1 + c_2\omega_3 + c_0\omega_3^2 \end{pmatrix} = \begin{pmatrix} (c_0 + c_1\omega_3 + c_2\omega_3^2)1\\ (c_0 + c_1\omega_3 + c_2\omega_3^2)\omega_3\\ (c_0 + c_1\omega_3 + c_2\omega_3^2)\omega_3^2 \end{pmatrix}.$$

We see that  $(1, \omega_3, \omega_3^2)^t$  is an eigenvector of C associated with the eigenvalue  $c_0 + c_1\omega_3 + c_2\omega_3^2$ .

More generally we have the following theorem.

### Theorem (The Circulant Diagonalization Theorem)

Let

$$C = \begin{bmatrix} c_0 & c_1 & \cdots & c_{n-2} & c_{n-1} \\ c_{n-1} & c_0 & c_1 & & c_{n-2} \\ \vdots & c_{n-1} & c_0 & \ddots & \vdots \\ c_2 & & \ddots & \ddots & c_1 \\ c_1 & c_2 & \cdots & c_{n-1} & c_0 \end{bmatrix}.$$

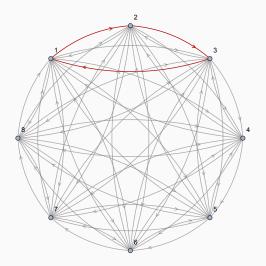
be the circulant matrix formed by the vector  $(c_0, c_1, \dots, c_{n-1})^T$ . Let

$$v_{n,j} = \left(1, \omega_n^j, \omega_n^{2j}, \dots, \omega_n^{(n-1)j}\right)^T, \quad j = 0, 1, \dots, n-1.$$

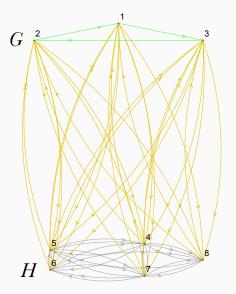
Then  $v_{n,j}$  is an eigenvector of C associated with the eigenvalue

$$\lambda_j^C = c_0 + c_1 \omega_n^j + c_2 \omega_n^{2j} + \dots + c_{n-1} \omega_n^{(n-1)j}$$

Originally, we are interested in graphs obtained by removing a directed cycle in a complete graph. Here is one example.



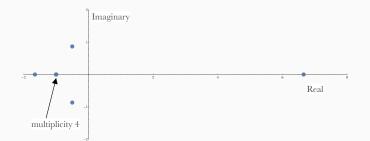
Here is another way to represent this graph



With this new representation, the adjacency matrix of this graph has the form

$$\begin{pmatrix} 0 & 0 & 1 & 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 0 & 1 & 1 & 1 & 1 & 1 \\ 0 & 1 & 0 & 1 & 1 & 1 & 1 & 1 \\ \hline 1 & 1 & 1 & 0 & 1 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 0 & 1 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 0 & 1 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 1 & 1 & 1 & 1 & 0 \end{pmatrix}$$

With this new representation, the adjacency matrix of this graph has the form



From this example, we are convinced that it would be quite interesting to study matrices of the following form

$$A = \begin{pmatrix} C & 1_{k_1,k_2} \\ 1_{k_2,k_1} & D \end{pmatrix},$$

with  $C = Circ(c_0, \ldots, c_{k_1-1})$  and  $D = Circ(d_0, \ldots, d_{k_2-1})$ . Here  $1_{m,n}$  is the matrix of size  $m \times n$  with all entries equal to 1.

Here is a crucial observation.

## **Proposition**

For  $1 \le j \le k_1 - 1$  let

$$w_j = (1, \omega_{k_1}^j, \omega_{k_1}^{2j}, \dots, \omega_{k_1}^{(k_1-1)j}, \underbrace{0 \dots, 0}_{k_2 \text{ zeros}})^T = v_{k_1,j} * \vec{0}_{k_2}.$$

Then  $w_j$  is an eigenvector of A associated with the eigenvalue

$$\lambda_j^C = c_0 + c_{k_1-1}\omega_{k_1}^j + c_{k_1-2}\omega_{k_1}^{2j} + \dots + c_1\omega_{k_1}^{(k_1-1)j}.$$

We have

$$Aw_{j} = \begin{pmatrix} C & 1_{k_{1},k_{2}} \\ 1_{k_{2},k_{1}} & D \end{pmatrix} \begin{pmatrix} v_{k_{1},j} \\ 0 \end{pmatrix} = \begin{pmatrix} Cv_{k_{1},j} \\ 1_{k_{2},k_{1}}v_{k_{1},j} \end{pmatrix}.$$

We have

$$Aw_{j} = \begin{pmatrix} C & 1_{k_{1},k_{2}} \\ 1_{k_{2},k_{1}} & D \end{pmatrix} \begin{pmatrix} v_{k_{1},j} \\ 0 \end{pmatrix} = \begin{pmatrix} Cv_{k_{1},j} \\ 1_{k_{2},k_{1}}v_{k_{1},j} \end{pmatrix}.$$

In other words,

$$Aw_j = Cv_{k_1,j} * \underbrace{\left(t_j, t_j, \dots, t_j\right)^T}_{k_2 \text{ terms}}. = \lambda_j^C v_{k_1,j} * \underbrace{\left(t_j, t_j, \dots, t_j\right)^T}_{n-k \text{ terms}}$$

where

$$t_j = \sum_{i=0}^{k_1-1} \omega_k^{ij}.$$

We have

$$Aw_{j} = \begin{pmatrix} C & 1_{k_{1},k_{2}} \\ 1_{k_{2},k_{1}} & D \end{pmatrix} \begin{pmatrix} v_{k_{1},j} \\ 0 \end{pmatrix} = \begin{pmatrix} Cv_{k_{1},j} \\ 1_{k_{2},k_{1}}v_{k_{1},j} \end{pmatrix}.$$

In other words,

$$Aw_j = Cv_{k_1,j} * \underbrace{\left(t_j, t_j, \dots, t_j\right)^T}_{k_2 \text{ terms}}. = \lambda_j^C v_{k_1,j} * \underbrace{\left(t_j, t_j, \dots, t_j\right)^T}_{n-k \text{ terms}}$$

where

$$t_j = \sum_{i=0}^{k_1-1} \omega_k^{ij}.$$

By the assumption  $1 \le j \le k_1 - 1$ , we have  $t_j = 0$ . Therefore,  $Aw_j = \lambda_j^C w_j$ .

We have

$$Aw_{j} = \begin{pmatrix} C & 1_{k_{1},k_{2}} \\ 1_{k_{2},k_{1}} & D \end{pmatrix} \begin{pmatrix} v_{k_{1},j} \\ 0 \end{pmatrix} = \begin{pmatrix} Cv_{k_{1},j} \\ 1_{k_{2},k_{1}}v_{k_{1},j} \end{pmatrix}.$$

In other words,

$$Aw_{j} = Cv_{k_{1},j} * \underbrace{(t_{j}, t_{j}, \dots, t_{j})^{T}}_{k_{2} \text{ terms}}. = \lambda_{j}^{C} v_{k_{1},j} * \underbrace{(t_{j}, t_{j}, \dots, t_{j})^{T}}_{n-k \text{ terms}}$$

where

$$t_j = \sum_{i=0}^{k_1-1} \omega_k^{ij}.$$

By the assumption  $1 \le j \le k_1 - 1$ , we have  $t_j = 0$ . Therefore,  $Aw_j = \lambda_j^C w_j$ . In other words,  $w_j$  is an eigenvector of A associated with the eigenvalue  $\lambda_i^C$  for  $1 \le j \le k_1 - 1$ .

By symmetry, we can also see that

## **Proposition**

For  $1 \le j \le k_2 - 1$ , let

$$z_j = (\underbrace{0, \dots, 0}_{k_1 \text{ zeros}}, 1, \omega_{k_2}^j, \omega_{k_2}^{2j}, \dots, \omega_{k_2}^{(k_2-1)j})^T = \vec{0}_{k_1} * v_{k_2,j}.$$

Then  $z_j$  is an eigenvector of A associated with the eigenvalue

$$\lambda_j^D = d_0 + d_1 \omega_{k_2}^j + d_2 \omega_{k_2}^{2j} + \dots + d_{k_2-1} \omega_{k_2}^{(k_2-1)j}$$

In summary, we have been able to find  $k_1+k_2-2$  eigenvectors of A. We need to find two more. We will look for eigenvectors of the form

$$v = (\underbrace{x, x, \dots, x}_{k_1 \text{ terms}}, \underbrace{y, y, \dots, y}_{k_2 \text{ terms}})^T$$

In summary, we have been able to find  $k_1 + k_2 - 2$  eigenvectors of A. We need to find two more. We will look for eigenvectors of the form

$$v = (\underbrace{x, x, \dots, x}_{k_1 \text{ terms}}, \underbrace{y, y, \dots, y}_{k_2 \text{ terms}})^T$$

By explicit calculations, we have

$$Av = (\underbrace{C_s x + k_2 y, \dots, C_s x + k_2 y}_{k_1 \text{ terms}}, \underbrace{k_1 x + D_s y, \dots, k_1 x + D_s y}_{k_2 \text{ terms}})^T.$$

Here 
$$C_s = \sum_{i=0}^{k_1-1} c_i, D_s = \sum_{i=0}^{k_2-1} d_i$$
.

In summary, we have been able to find  $k_1 + k_2 - 2$  eigenvectors of A. We need to find two more. We will look for eigenvectors of the form

$$v = (\underbrace{x, x, \dots, x}_{k_1 \text{ terms}}, \underbrace{y, y, \dots, y}_{k_2 \text{ terms}})^T$$

By explicit calculations, we have

$$Av = (\underbrace{C_s x + k_2 y, \dots, C_s x + k_2 y}_{k_1 \text{ terms}}, \underbrace{k_1 x + D_s y, \dots, k_1 x + D_s y}_{k_2 \text{ terms}})^T.$$

Here  $C_s = \sum_{i=0}^{k_1-1} c_i$ ,  $D_s = \sum_{i=0}^{k_2-1} d_i$ . The condition  $Av = \lambda v$  is equivalent to

$$\begin{pmatrix} C_s & k_2 \\ k_1 & D_s \end{pmatrix} \begin{pmatrix} x \\ y \end{pmatrix} = \lambda \begin{pmatrix} x \\ y \end{pmatrix}.$$

In other words, v is an eigenvector of A if and only if  $(x, y)^T$  is an eigenvector of

$$\bar{A} = \begin{pmatrix} C_s & k_2 \\ k_1 & D_s \end{pmatrix}.$$

### The main theorem

In summary, we have the following theorem.

### **Theorem**

Let  $\{(x_1, y_1)^T, (x_2, y_2)^T\}$  be an (generalized) eigenbasis for  $\bar{A}$ . Let

$$v_1 = \underbrace{(x_1, x_1, \dots, x_1)}_{k_1 \text{ terms}}, \underbrace{y_1, y_1, \dots, y_1}_{k_2 \text{ terms}})^T,$$

$$v_2 = \underbrace{(x_2, x_2, \dots, x_2)}_{y_2, y_2, \dots, y_2}, \underbrace{y_2, y_2, \dots, y_2}_{y_2, y_2, \dots, y_2})^T.$$

ko terms

k<sub>1</sub> terms

Then the system  $\{\omega_j\}_{j=1}^{k_1-1} \cup \{z_j\}_{j=1}^{k_2-1} \cup \{v_1, v_2\}$  of eigenvectors of A is linearly independent. In other words, A is diagonalizable by these eigenvectors.

### **Further results**

The above approach can be generalized to study the join of several circulant matrices

$$A = \begin{pmatrix} C_1 & a_{1,2}1 & \cdots & a_{1,d}1 \\ \hline a_{2,1}1 & C_2 & \cdots & a_{2,d}1 \\ \hline \vdots & \vdots & \ddots & \vdots \\ \hline a_{d,1}1 & a_{d,2}1 & \cdots & C_d \end{pmatrix}.$$

Here  $C_i$  is a circulant matrix of size  $k_i \times k_i$  for each  $1 \le i \le d$ , and  $a_{i,j}1$  is a  $k_i \times k_j$  matrix with all entries equal to a constant  $a_{i,j}$ .

#### **Further results**

The above approach can be generalized to study the join of several circulant matrices

$$A = \begin{pmatrix} C_1 & a_{1,2}1 & \cdots & a_{1,d}1 \\ \hline a_{2,1}1 & C_2 & \cdots & a_{2,d}1 \\ \hline \vdots & \vdots & \ddots & \vdots \\ \hline a_{d,1}1 & a_{d,2}1 & \cdots & C_d \end{pmatrix}.$$

Here  $C_i$  is a circulant matrix of size  $k_i \times k_i$  for each  $1 \le i \le d$ , and  $a_{i,j}1$  is a  $k_i \times k_j$  matrix with all entries equal to a constant  $a_{i,j}$ . If we just care about the spectrum of A, we only need to assume that  $C_i$  is normal with constant row sums.

# Thank you

