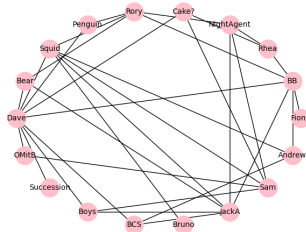


Lecture 7: Permutations and Bipartite Networks

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These slides are by Niall Madden. Elements are based on "A First Course in Network Theory" by Estrada and Knight. Also AC's notes...

Outline

- | | | |
|---|-----------------------------------|----------------------------|
| 1 | Thanks for completing the survey! | ■ Connected graphs |
| 2 | Graph Connectivity | 4 Connected Components |
| 3 | Permutation matrices | 5 Bipartite Graphs (again) |
| | | 6 Exercise(s) |

For further reading, see Section 2.4 of [A First Course in Network Theory \(Knight\)](#).

Slides are at:

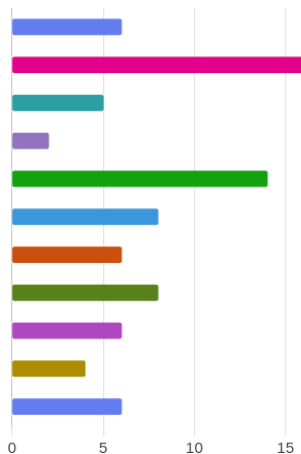
<https://www.niallmadden.ie/2425-CS4423>



Thanks for completing the survey!

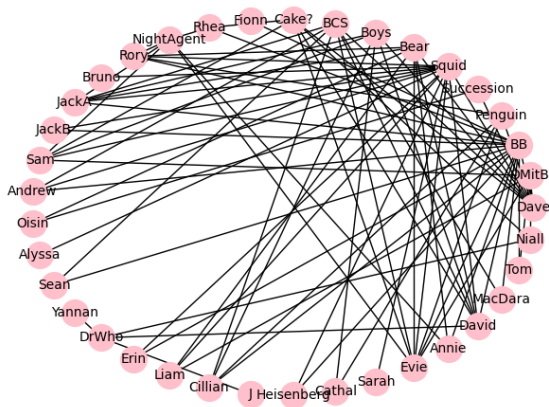
Here is some of the data we collected:

● Only Murders in the Building	6
● Breaking Bad	16
● The Penguin	5
● Succession	2
● Squid Game	14
● The Bear	8
● The Boys	6
● Better Call Saul	8
● Night Agent	6
● Dr Who	4
● Is it Cake?	6



Thanks for completing the survey!

Here is what it looks like as a graph:



Its order is 37, and size is 81; we'll return to this later...

Graph Connectivity

- ▶ A graph/network is **connected** if there is a path between every pair of nodes.
- ▶ If the graph is *not* connected, we say it is **disconnected**.
- ▶ We now know how to check if a graph is connected by looking at powers of its adjacency matrix. However, that is not very practical for large networks.
- ▶ However, we can determine if a graph is connected, but just looking at the adjacency matrix, providing we have ordered the nodes properly.

Permutation matrices

We know that the structure of a network is not changes by relabelling its nodes. Sometimes, it is is useful to relabel them in order to expose certain properties, such as connectivity.

Example:

Since we think of the nodes as all being numbered from 1 to n , this is the same as **permuting** the numbers of some subset of the nodes.

Permutation matrices

When working with the adjacency matrix of a graph, such a permutation is expressed in terms of a **permutation matrix**, P : this is a $0 - 1$ matrix (a.k.a. a “Boolean” or “binary” matrix), where there is a single 1 on every row and column.

If the nodes of a graph G (with adjacency matrix A) are listed as entries in a vector, q , then

- ▶ Pq is a permutation of the nodes, and
- ▶ PAP^T is the adjacency matrix of the graph with that node permutation applied.

Permutation matrices are important when studying graph connectivity because...

FACT!

A graph with adjacency matrix A is **disconnected** if and only if there is a permutation matrix P such that

$$A = P \begin{pmatrix} X & O \\ O^T & Y \end{pmatrix},$$

where O represents the zero matrix with the same number of rows as X and the same number of columns as Y .

Example:

Connected Components

If a network is not connected, then we can divide it into **components** which *are* connected.

The number of connected components is the number of blocks in the permuted adjacency matrix:

Bipartite Graphs (again)

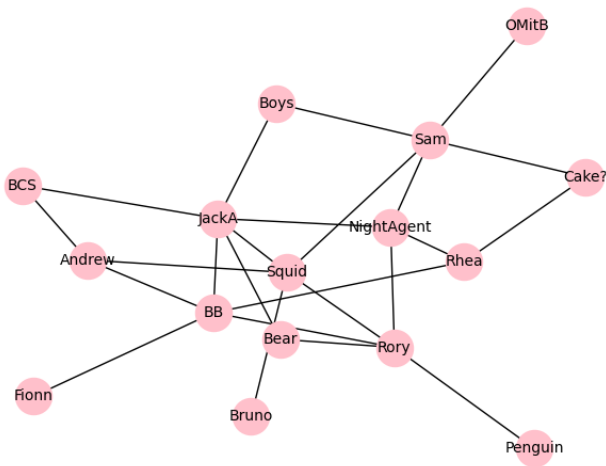
One reason we did the survey is that the resulting data set is a good example of a **bipartite** graph: nodes represent either people or programmes that they watch, with an edge between a person and a programme that they watch.

So the graph must be bipartite.

Such a graph is called an **affiliation** network;

Bipartite Graphs (again)

Here is a **subgraph** of our survey, of order 16 and size 24, based on 7 randomly chosen people:



Bipartite Graphs (again)

This is the adjacency matrix:

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

Bipartite Graphs (again)

That version of the adjacency matrix is not very insightful. But ordering the nodes so that people are listed first we get the matrix:

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

Bipartite Graphs (again)

Let's consider $B = A^2$:

$$B = \begin{pmatrix} 1 & 1 & 1 & 0 & 1 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 3 & 2 & 0 & 2 & 2 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 2 & 5 & 1 & 4 & 2 & 2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 1 & 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 2 & 4 & 1 & 6 & 3 & 3 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 2 & 2 & 1 & 3 & 5 & 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1 & 1 & 2 & 1 & 3 & 1 & 3 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 5 & 3 & 1 & 1 & 3 & 2 & 1 & 2 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 3 & 4 & 2 & 1 & 3 & 2 & 2 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 2 & 2 & 0 & 1 & 0 & 1 & 0 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 1 & 1 & 1 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 3 & 3 & 1 & 1 & 5 & 2 & 2 & 2 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 & 2 & 0 & 1 & 2 & 2 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 2 & 1 & 0 & 2 & 1 & 2 & 1 & 1 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 2 & 1 & 0 & 0 & 2 & 1 & 1 & 2 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 1 & 0 & 1 & 0 & 1 & 0 & 1 \end{pmatrix}$$

Bipartite Graphs (again)

We finished here on Wednesday, but will take up at the same point on Thursday.

Exercise(s)

1. Let u be a vector with n entries. Let $D = \text{diag}(u)$. That is, $D = (d_{ij})$ is the diagonal matrix with entries

$$d_{ij} = \begin{cases} u_i & i = j \\ 0 & i \neq j. \end{cases}$$

Verify that $PDP^T = \text{diag}(Pu)$.

2. In all the examples we looked at, we had a symmetric P . Is every permutation matrix symmetric? If so, explain why. If not, give an example.