

December 1st, 2023

1 Some Graph Theory

We can consider a simple graph network to help facilitate the proof:

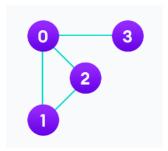


Figure 1: A simple undirected graph

This graph has a corresponding adjacent matrix:

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

We can prove by induction that in an unweighted undirected graph, the (i, j) entry of the m^{th} power of adjacency matrix A counts the number of walks of length m from i to j.

We begin with m=2, and consider nodes 1 and 2, i=1, j=2. Visually, we can perceive that there is only 1 walk between node 1 and node 2 than can be done in 2 lengths. This is the walk $1 \to 0 \to 2$.

$$A^{2} = \begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & \mathbf{1} & 0 \\ 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{bmatrix}^{2} = \begin{bmatrix} 3 & 1 & 1 & 0 \\ 1 & 2 & \mathbf{1} & 1 \\ 1 & 1 & 2 & 1 \\ 0 & 1 & 1 & 1 \end{bmatrix}$$

Clearly, this is correct. Similarly, we can see the assumption is correct for all other nodes as well. We now suppose that the assumption is still correct for m = k. We can show that it will hold for m = k + 1.

With m = k, we have that the number of walks is $(A_{ij})^k$. When m = k + 1, we can consider a third vertex, p, which is a 1 length walk from j. Thus, the number of walks from i to j of length k + 1 is the sum of the number of walks from i to j of length k multiplied by the number of walks from i to i of length i.

Thus,

$$(A_{ij})^{k+1} = \sum_{p \in V(G)} (A_{ip})^k (A_{pj})$$
$$LS = RS$$

By induction, we have proven that the (i, j) entry of the m^{th} power of adjacency matrix A counts the number of walks of length m from i to j.

2 Random Walks

A simple random walk with the following transition probabilities is given by:

$$P_{ij} = Pr(X_{k+1} = j | X_k = i) = \begin{cases} \frac{1}{d(i)} & \text{if } ij \in E \\ 0 & \text{otherwise} \end{cases}$$

where d(i) is the degree of node i, or how many neighbours it has. A stationary distribution (row vector) of a random walk is a distribution π such that $\pi P = \pi$.

If ρ_k is a row vector giving the probability distribution of X_k ,

$$\rho_k(i) = P(X_k = i), i \in V(G)$$
$$\rho_{k+1} = \rho_k P$$

For a walk of length k, we have that $\rho_k = \rho_0 P^k$, where ρ_0 is the initial distribution.

The Perron-Frobenius theorem implies the existence of a stationary distribution π which is a positive left eigenvector of P corresponding to the eigenvalue 1. This means that:

$$\pi P = \pi$$

$$\pi(i) > 0$$

$$\sum_{i \in V(G)} \pi(i) = 1$$

If the initial vertex of the walk is chosen according to π , then the distribution at time k is also π . Hence,

$$\rho_k = \pi P^k = \pi$$

$$P(X_k = i) = \pi(i)$$

To find a probability distribution π that satisfies the above conditions, we solve:

$$\pi(i)p_{ij} = \pi(j)p_{ji}$$

And when this condition is satisfied:

$$\pi(i) = \sum_{j \in V(G)} \pi(i) p_{ij} = \sum_{i \in V(G)} \pi(j) p_{ji}$$

With the transition probabilities given in the question, we have:

$$\frac{\pi(i)}{d(i)} = \frac{\pi(j)}{d(j)} = k$$

Exploring further:

$$1 = \sum_{i \in j} \pi(i) = k \sum_{i \in V(G)} d(i) = k(2m)$$

Thus,

$$\pi(i) = kd(i)$$

$$\pi(i) = \frac{d(i)}{2m} = \frac{d(i)}{n}$$

The stationary distribution for each node is clearly proportional to its degree.

3 Random Walks without a Restart