



# Random walk with barriers on a graph

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## Abstract

We obtain expected number of arrivals, absorption probabilities and expected time until absorption for an asymmetric discrete random walk on a graph in the presence of multiple function barriers. On each edge of the graph and in each vertex (barrier) specific probabilities are defined.

**Keywords:** random walk, absorption, graph.

**msc:** 60G50, 60J05.

## 1 Introduction

Random walk can be used in various disciplines: in economics to model share prices and their derivatives, in medicine and biology where absorbing barriers give a natural model for a wide variety of phenomena, in physics as a simplified model of Brownian motion, in ecology to describe individual animal movements and population dynamics, in statistics to analyze sequential test procedures, in computer science to estimate the size of the World Wide Web using randomized algorithms. Burioni and Cassi (2005) give a review of random walks on graphs, where the generalization of the concept of dimension to inhomogeneous structures, using infinite graphs, is considered. Durhuus, Jonsson, and Wheeler (2006) develop techniques to obtain rigorous bounds on the behavior of random walks on combs. Using these bounds they calculate the spectral dimension of random combs with infinite teeth at random positions or teeth with random but finite length. Random walks have been studied for decades on regular structures such as lattices. We now give a brief historical review of the use of barriers in a one-dimensional discrete random walk. Weesakul (1961) discussed the classical problem of random walk restricted between a reflecting and an absorbing barrier. Using generating functions he obtains explicit expressions for the probability of absorption. Lehner (1963) studies one-dimensional random walk with a partially reflecting barrier using combinatorial methods. Gupta (1966) introduces the concept of a multiple function barrier (MFB): a state that can absorb, reflect, let through or hold for a moment. Dua, Khadilkar, and Sen (1976) find the bivariate generating functions of the

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probabilities of a particle reaching a certain state under different conditions. Percus (1985) considers an asymmetric random walk, with one or two boundaries, on a one-dimensional lattice. At the boundaries, the walker is either absorbed or reflected back to the system. Using generating functions the probability distribution of being at position  $m$  after  $n$  steps is obtained, as well as the mean number of steps until absorption. El-Shehawey (2000) obtains absorption probabilities at the boundaries for a random walk between one or two partially absorbing boundaries as well as the conditional mean for the number of steps before stopping given the absorption at a specified barrier, using conditional probabilities. In this paper we obtain expected number of arrivals, absorption probabilities and expected time until absorption for an asymmetric discrete random walk with multiple function barriers. Our graph consists of multiple function barriers (vertices) and states on the edges between the MFB's. On each edge of the graph a random walk with its own states and jumping probabilities is introduced. When the walker reaches a multiple function barrier a random process is activated according to a set of probabilities, or the particle is absorbed in the barrier. Each barrier has its own probability parameters. In section 2 we use generating functions to find the expected number of arrivals to any state, the probability of absorption and the expected time until absorption. In section 3 we analyze some examples of graphs with multiple function barriers: a star graph and a cycle graph.

## 2 A graph with multiple function barriers

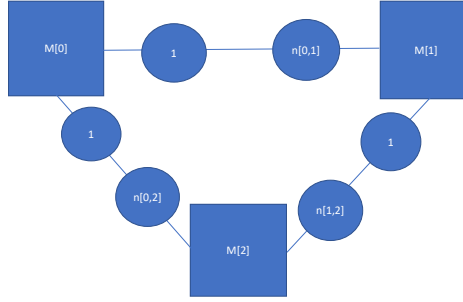
### 2.1 Description of the random walk

In a finite graph we have vertices  $M[0], M[1], \dots, M[N]$  representing the MFB's. Between  $M[i]$  and  $M[j]$  there is a random walk with a finite number of states  $n[i, j]$ , which we number  $1, 2, \dots, n[i, j]$  in the direction from  $M[i]$  to  $M[j]$  when  $i < j$ . Example 1. Random walk on an interval with two reflecting/absorbing barriers:



Example 2. Random walk on a triangle:

## 2. A graph with multiple function barriers



We will use the abbreviation  $[i, j]$  for the edge between  $M[i]$  and  $M[j]$ . Each random walk from  $M[i]$  to  $M[j]$  has its own parameters  $p = p[i, j]$  and  $q = q[i, j]$ , where  $p$  is the one-step forward probability and  $q$  one-step backward probability ( $p + q = 1$ ). We demand  $p[i, j].q[i, j] > 0$  for each  $i$  and  $j$ . In  $M[i]$  there is probability  $p_{[i, j]}^*$  to move one step in the direction of  $M[j]$  ( $0 \leq i, j \leq N$ ) and probability  $p_{[i, i]}^*$  for absorption in  $M[i]$  ( $0 \leq i \leq N$ ), where  $\sum_{j=0}^N p_{[i, j]}^* = 1$  ( $i = 0, 1, \dots, N$ ). We start in  $M[0]$ .

### 2.2 Expected number of arrivals

We are interested in the expected number of arrivals in the MFB's as well as the expected number of arrivals in the other states of the graph. Let  $p_{ij}^{(m)}$  be the probability that the system is in state  $j$  after  $m$  steps when starting in  $i$ . If  $j$  is not a MFB:

$$X_j = X_j(z) = X_{i,j}(z) = \sum_{m=0}^{\infty} p_{ij}^{(m)} z^m$$

Expected number of arrivals in  $j$  when starting in  $i$ :

$$x_j = x_{i,j} = X_j(1)$$

For MFB  $M[j]$ :

$$Y_j = Y_j(z) = Y_{i,M[j]}(z) = \sum_{m=0}^{\infty} p_{i,M[j]}^{(m)} z^m$$

Expected number of arrivals in  $M[j]$  when starting in  $i$ :

$$y_j = y_{i,j} = Y_j(1)$$

On edge  $[i, j]$ :

$$\rho = \rho[i, j] = \frac{p[i, j]}{q[i, j]}; \quad n = n[i, j]$$

**Theorem 1** –  $y_k$  ( $k = 0, 1, \dots, N$ ) is the unique solution of  $\sum_{j=0}^N u_{ij}y_j = -\delta(i, 0)$  ( $i = 0, 1 \dots N$ ) where

$$u_{ij} = \left[ \frac{(1-\rho)\rho^n}{1-\rho^{n+1}} \right] p_{[j,i]}^* \quad (j < i, \rho \neq 1)$$

$$u_{ij} = \left[ \frac{1-\rho}{1-\rho^{n+1}} \right] p_{[j,i]}^* \quad (j > i, \rho \neq 1)$$

$$u_{ij} = \frac{p_{[j,i]}^*}{n+1} \quad (j \neq i, \rho = 1)$$

$$u_{ii} = -1 + \sum_{j < i, \rho \neq 1} \left[ \frac{\rho(1-\rho^n)}{1-\rho^{n+1}} \right] p_{[i,j]}^* + \sum_{j > i, \rho \neq 1} \left[ \frac{1-\rho^n}{1-\rho^{n+1}} \right] p_{[i,j]}^* + \sum_{j \neq i, \rho = 1} \left[ \frac{n}{n+1} \right] p_{[i,j]}^*$$

*Proof.* CASE 1: ( $z \neq 1$ )  $\vee$  ( $p \neq q$ ). We prove the results of CASE 1 in 5 steps.

Step 1: The random walk between  $M[i]$  and  $M[j]$  ( $0 \leq i < j \leq N$ ) is described by considering the last step of the random walk:

$$X_k = pzX_{k-1} + qzX_{k+1} \quad (1)$$

where  $X_k = X_k^{[i,j]}$ ,  $p = p[i, j]$ ,  $q = q[i, j]$ . Characteristic equation:

$$qz\lambda^2 - \lambda + pz = 0 \quad (2)$$

with solutions  $\lambda_1$  and  $\lambda_2$  with  $\lambda_1 > \lambda_2$  and  $\lambda_1\lambda_2 = \frac{p}{q} = \rho$ . So:

$$X_k = a\lambda_1^k + b\lambda_2^k \quad (\lambda_1 > \lambda_2) \quad (3)$$

Step 2: We express  $a$  and  $b$  in  $Y_i$  and  $Y_j$ . Focus on states 1 and  $n = n[i, j]$  between  $M[i]$  and  $M[j]$  ( $0 \leq i < j \leq N$ ) and their neighbors. Considering the last step of the random walk we get:

$$X_1 = p_{[i,j]}^* z Y_i + qz X_2 \quad (4)$$

$$X_n = pz X_{n-1} + p_{[j,i]}^* z Y_j \quad (5)$$

Using (3), (4) and (5) we get:

$$(\lambda_2^{n+1} - \lambda_1^{n+1})a = \lambda_2^{n+1} \frac{p_{[i,j]}^*}{p} Y_i - \frac{p_{[j,i]}^*}{q} Y_j \quad (6)$$

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$$(\lambda_2^{n+1} - \lambda_1^{n+1})b = \frac{P_{[j,i]}^*}{q} Y_j - \lambda_1^{n+1} \frac{P_{[i,j]}^*}{p} Y_i \quad (7)$$

Step 3: We express  $X_1$  and  $X_n$  in  $Y_i$  and  $Y_j$ . Using (3) with  $k = 1$  and  $k = n$  in combination with (6) and (7) gives:

$$q(\lambda_2^{n+1} - \lambda_1^{n+1})X_1 = (\lambda_2 - \lambda_1)p_{j,i}^* Y_j + (\lambda_2^n - \lambda_1^n)p_{i,j}^* Y_i \quad (i < j) \quad (8)$$

$$p(\lambda_2^{n+1} - \lambda_1^{n+1})X_n = (\lambda_2 - \lambda_1)(\lambda_1 \lambda_2)^n p_{i,j}^* Y_i + \lambda_1 \lambda_2 (\lambda_2^n - \lambda_1^n) p_{j,i}^* Y_j \quad (i < j)$$

We need the last formula for  $j < i$ , so we interchange  $i$  and  $j$ :

$$p(\lambda_2^{n+1} - \lambda_1^{n+1})X_n = (\lambda_2 - \lambda_1)(\lambda_1 \lambda_2)^n p_{j,i}^* Y_j + \lambda_1 \lambda_2 (\lambda_2^n - \lambda_1^n) p_{i,j}^* Y_i \quad (j < i) \quad (9)$$

Step 4: Focus on  $M[i]$  and its neighbors  $X_1[i, j]$  ( $j > i$ ) and  $X_n[i, j]$  ( $j < i$ ). Using (8) and (9) we get, considering the last step of the random walk:

$$\begin{aligned} Y_i &= \sum_{j>i} q[i, j] z X_1[i, j] + \sum_{j<i} p[i, j] z X_n[i, j] + \delta(i, 0) = \\ & z \sum_{j>i} (\lambda_2^{n+1} - \lambda_1^{n+1})^{-1} \left[ (\lambda_2 - \lambda_1) p_{j,i}^* Y_j + (\lambda_2^n - \lambda_1^n) p_{i,j}^* Y_i \right] + \\ & z \sum_{j<i} (\lambda_2^{n+1} - \lambda_1^{n+1})^{-1} \left[ (\lambda_2 - \lambda_1) (\lambda_1 \lambda_2)^n p_{j,i}^* Y_j + \lambda_1 \lambda_2 (\lambda_2^n - \lambda_1^n) p_{i,j}^* Y_i \right] + \delta(i, 0) = \\ & z Y_i \left\{ \sum_{j>i} \left[ \frac{\lambda_2^n - \lambda_1^n}{\lambda_2^{n+1} - \lambda_1^{n+1}} \right] p_{[i,j]}^* + \sum_{j<i} \left[ \frac{\lambda_1 \lambda_2 (\lambda_2^n - \lambda_1^n)}{\lambda_2^{n+1} - \lambda_1^{n+1}} \right] p_{[i,j]}^* \right\} + \\ & z Y_j \left\{ \sum_{j>i} \left[ \frac{\lambda_2 - \lambda_1}{\lambda_2^{n+1} - \lambda_1^{n+1}} \right] p_{[j,i]}^* + \sum_{j<i} \left[ \frac{(\lambda_2 - \lambda_1) (\lambda_1 \lambda_2)^n}{\lambda_2^{n+1} - \lambda_1^{n+1}} \right] p_{[j,i]}^* \right\} + \delta(i, 0) \end{aligned}$$

Step 5: If  $(z = 1) \wedge (\rho > 1)$  then  $\lambda_1 = \rho; \lambda_2 = 1$ . If  $(z = 1) \wedge (\rho < 1)$  then  $\lambda_1 = 1; \lambda_2 = \rho$ . We get the result by observing the coefficients of  $y_i$  and  $y_j$  and the constants.

CASE 2:  $(z = 1) \wedge (p = q)$

We can use the same method as in CASE 1, but now with  $x$  and  $y$  instead of  $X$  and  $Y$ , starting with  $x_k = ak + b$ , but we prefer a faster way by applying l'Hospitals rule in the asymmetric case:

$$\begin{aligned} \lim_{\rho \rightarrow 1} \frac{(1 - \rho)\rho^n}{1 - \rho^{n+1}} &= \frac{1}{n+1} = \lim_{\rho \rightarrow 1} \frac{1 - \rho}{1 - \rho^{n+1}} \\ \lim_{\rho \rightarrow 1} \frac{\rho(1 - \rho^n)}{1 - \rho^{n+1}} &= \frac{n}{n+1} = \lim_{\rho \rightarrow 1} \frac{1 - \rho^n}{1 - \rho^{n+1}} \end{aligned} \quad \square$$

**Theorem 2** – Case  $\rho \neq 1$ :

$$x_k = \frac{(1 - \rho^k) \frac{p_{[j,i]}^*}{q} y_j + (\rho^k - \rho^{n+1}) \frac{p_{[i,j]}^*}{p} y_i}{1 - \rho^{n+1}} \quad (10)$$

Case  $\rho = 1$ :

$$x_k = \frac{kp_{[j,i]}^* y_j + (n+1-k)p_{[i,j]}^* y_i}{2(n+1)} \quad (11)$$

*Proof.* Case  $\rho \neq 1$ : Use (3), (6) and (7) with  $z = 1$  and  $\lambda_1 = \rho; \lambda_2 = 1(\rho > 1)$  or  $\lambda_1 = 1; \lambda_2 = \rho(\rho < 1)$ . The result of  $\rho = 1$  is obtained by applying l'Hospitals rule for the asymmetric case.  $\square$

**Theorem 3** –

$$\sum_{j=0}^N p_{[j,j]}^* y_j = 1 \quad (12)$$

*Proof.* Consider first  $\rho \neq 1$ . Using Theorem 1 we get:

$$\begin{aligned} \sum_{i=0}^N u_{ij} &= u_{jj} + \sum_{i < j} u_{ij} + \sum_{i > j} u_{ij} = -1 + \sum_{j < i} \left[ \frac{\rho(1 - \rho^n)}{1 - \rho^{n+1}} \right] p_{[i,j]}^* + \\ &\quad \sum_{j > i} \left[ \frac{1 - \rho^n}{1 - \rho^{n+1}} \right] p_{[i,j]}^* + \sum_{i < j} \left[ \frac{1 - \rho}{1 - \rho^{n+1}} \right] p_{[j,i]}^* + \sum_{i > j} \left[ \frac{(1 - \rho)\rho^n}{1 - \rho^{n+1}} \right] p_{[j,i]}^* \end{aligned}$$

Interchange  $i$  and  $j$  in the first two summations and add terms with  $\sum_{i < j}$  and  $\sum_{i > j}$ :

$$\sum_{i=0}^N u_{ij} = -1 + \sum_{i \neq j} p_{[j,i]}^* = -p_{[j,j]}^* \quad (13)$$

We also have (Theorem 1)  $\sum_{j=0}^N u_{ij} y_j = -\delta(i, 0)$  ( $i = 0, 1 \dots N$ ), so  $\sum_{i=0}^N \sum_{j=0}^N u_{ij} y_j = -1$ . Using (13) we get:  $\sum_{i=0}^N \sum_{j=0}^N u_{ij} y_j = \sum_{j=0}^N \left[ \sum_{i=0}^N u_{ij} \right] y_j = -\sum_{j=0}^N p_{[j,j]}^* y_j$ . The symmetric case  $\rho = 1$  proceeds along the same lines.  $\square$

### 2.3 Expected time until absorption

Let  $t_k$  be the expected time until absorption when starting in  $M[k]$  ( $k = 0, 1, \dots, N$ ).

## 2. A graph with multiple function barriers

**Theorem 4**–  $t_k$  ( $k = 0, 1, \dots, N$ ) is the unique solution of  $\sum_{j=0}^N v_{ij} t_j = \tau_i$  ( $i = 0, 1, \dots, N$ ) where

$$v_{ij} = \left[ \frac{(1-\rho)\rho^n}{1-\rho^{n+1}} \right] p_{[i,j]}^* \quad (j > i, \rho \neq 1)$$

$$v_{ij} = \left[ \frac{1-\rho}{1-\rho^{n+1}} \right] p_{[i,j]}^* \quad (j < i, \rho \neq 1)$$

$$v_{ij} = \frac{p_{[i,j]}^*}{n+1} \quad (j \neq i, \rho = 1)$$

$$v_{ii} = -1 + \sum_{j < i, \rho \neq 1} \left[ \frac{\rho(1-\rho^n)}{1-\rho^{n+1}} \right] p_{[i,j]}^* + \sum_{j > i, \rho \neq 1} \left[ \frac{1-\rho^n}{1-\rho^{n+1}} \right] p_{[i,j]}^* + \sum_{j \neq i, \rho = 1} \left[ \frac{n}{n+1} \right] p_{[i,j]}^*$$

$$\begin{aligned} \tau_i = -1 + \sum_{j < i, \rho \neq 1} \left[ \frac{n-(n+1)\rho + \rho^{n+1}}{(p-q)(1-\rho^{n+1})} \right] p_{[i,j]}^* + \\ \sum_{j > i, \rho \neq 1} \left[ \frac{1-(n+1)\rho^n + n\rho^{n+1}}{(p-q)(1-\rho^{n+1})} \right] p_{[i,j]}^* - \sum_{j \neq i, \rho = 1} n p_{[i,j]}^* \end{aligned}$$

*Proof.* Step 1: Let  $m_k = m_k[i, j]$  ( $k = 1, 2, \dots, n[i, j]$ ) be the expected time until absorption when starting on edge  $[i, j]$  in state  $k$  ( $k = 1, 2, \dots, n[i, j]$ ). We have, considering the next step in the random walk:

$$m_k = p(m_{k+1} + 1) + q(m_{k-1} + 1) = pm_{k+1} + qm_{k-1} + 1 \quad (k = 1, 2, \dots, n[i, j])$$

with general solution (case  $\rho \neq 1$ ):

$$m_k = a\rho^{-k} + b - \frac{k}{p-q} \quad (k = 0, 1, \dots, n[i, j] + 1) \quad (14)$$

Step 2: We express  $a$  and  $b$  in  $t_i$  and  $t_j$  using (14) with  $k = 0$  and  $k = n+1$ :  $m_0 = a+b = t_i$  and  $m_{n+1} = a\rho^{-n-1} + b - \frac{n+1}{p-q} = t_j$  gives  $(1-\rho^{n+1})a = -\rho^{n+1}t_i + \rho^{n+1}t_j + \rho^{n+1}\left(\frac{n+1}{p-q}\right)$  and  $(1-\rho^{n+1})b = t_i - \rho^{n+1}t_j - \rho^{n+1}\left(\frac{n+1}{p-q}\right)$

Step 3: Using the expressions for  $a$  and  $b$  we get, using (14):

$$m_1 = \frac{(1-\rho^n)t_i + (\rho^n - \rho^{n+1})t_j}{1-\rho^{n+1}} + \frac{(n+1)(\rho^n - \rho^{n+1})}{(p-q)(1-\rho^{n+1})} - \frac{1}{p-q} \quad (i < j)$$

$$m_n = \frac{(1-\rho)t_i + (\rho - \rho^{n+1})t_j}{1-\rho^{n+1}} + \frac{(n+1)(\rho - \rho^{n+1})}{(p-q)(1-\rho^{n+1})} - \frac{n}{p-q} \quad (i < j)$$

We need the last formula with  $i > j$ , so by interchanging  $i$  and  $j$  we get:

$$m_n = \frac{(1-\rho)t_j + (\rho - \rho^{n+1})t_i}{1 - \rho^{n+1}} + \frac{(n+1)(\rho - \rho^{n+1})}{(p-q)(1 - \rho^{n+1})} - \frac{n}{p-q} \quad (i > j)$$

Step 4. Consider  $\rho = 1$  on edge  $[i, j]$ :  $m_k = \frac{1}{2}m_{k-1} + \frac{1}{2}m_{k+1} + 1$  gives

$$m_k = ak + b - k^2 \quad (k = 0, 1, \dots, n[i, j] + 1)$$

Along the same lines as in case  $\rho \neq 1$  we get:

$$m_1 = \frac{nt_i + t_j}{n+1} + n \quad (i < j)$$

$$m_n = \frac{nt_i + t_j}{n+1} + n \quad (i > j)$$

The same formula are found by applying l'Hospitals rule twice in case of  $\rho \neq 1$ .

Step 5: Substituting the values of  $m_1$  and  $m_n$  and considering the next step of the random walk:

$$\begin{aligned} t_i &= \sum_{j>i} p_{[i,j]}^* (m_1[i, j] + 1) + \sum_{j<i} p_{[i,j]}^* (m_n[i, j] + 1) + p_{[i,i]}^* \cdot 1 = \\ &1 + \sum_{j>i} p_{[i,j]}^* m_1[i, j] + \sum_{j<i} p_{[i,j]}^* m_n[i, j] = \\ &1 + \sum_{j>i, \rho \neq 1} p_{[i,j]}^* \left[ \frac{(1-\rho^n)t_i + (\rho^n - \rho^{n+1})t_j}{1 - \rho^{n+1}} + \frac{(n+1)(\rho^n - \rho^{n+1})}{(p-q)(1 - \rho^{n+1})} - \frac{1}{p-q} \right] + \\ &\sum_{j<i, \rho \neq 1} p_{[i,j]}^* \left[ \frac{(1-\rho)t_j + (\rho - \rho^{n+1})t_i}{1 - \rho^{n+1}} + \frac{(n+1)(\rho - \rho^{n+1})}{(p-q)(1 - \rho^{n+1})} - \frac{n}{p-q} \right] + \\ &\sum_{j>i, \rho=1} p_{[i,j]}^* \left[ \frac{nt_i + t_j}{n+1} + n \right] + \sum_{j<i, \rho=1} p_{[i,j]}^* \left[ \frac{nt_i + t_j}{n+1} + n \right] \end{aligned}$$

where  $i = 0, 1, \dots, N$ . By observing the coefficients of  $t_i$  and  $t_j$  and the constants we obtain the result.  $\square$

### 3 Examples of multiple function barrier graphs

#### 3.1 A finite star graph

We consider a star graph where the starting state is in the center  $M[0]$ . In a finite star graph we demand:

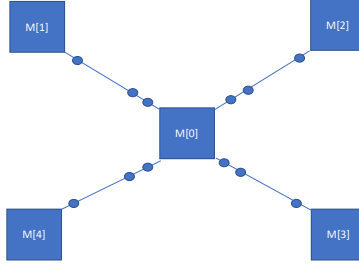
$$\sum_{i=0}^N p_{[0,i]}^* = 1 \quad \prod_{i=1}^N p_{[0,i]}^* \neq 0$$



### 3. Examples of multiple function barrier graphs

$$p_{[i,0]}^* + p_{[i,i]}^* = 1 \quad (i = 1, 2, \dots, N)$$

Example with  $N = 4$ :



We use the notation:  $p_i = p[0, i]$ ,  $q_i = q[0, i]$ ,  $\rho_i = \frac{p_i}{q_i}$  ( $i = 1, 2, \dots, N$ )

#### Expected number of arrivals

To obtain the expected number of arrivals in the finite star graph, we use Theorem 1 :  $\sum_{j=0}^N u_{ij} y_j = -\delta(i, 0)$  ( $i = 0, 1 \dots N$ ). We get (case  $\rho \neq 1$ ):

$$y_i = \left[ \begin{array}{c} -u_{i0} \\ u_{ii} \end{array} \right] y_0 = \left[ \frac{(1 - \rho_i) \rho_i^n p_{[0,i]}^*}{(1 - \rho_i) + \rho_i (1 - \rho_i^n) p_{[i,i]}^*} \right] y_0 \quad (i = 1, 2, \dots, N) \quad (15)$$

When  $\rho = 1$  we get:

$$y_i = \frac{p_{[0,i]}^*}{1 + n_i p_{[i,i]}^*} y_0 \quad (i = 1, 2, \dots, N) \quad (16)$$

Instead of the first equation of  $\sum_{j=0}^N u_{ij} y_j = -\delta(i, 0)$  ( $i = 0, 1 \dots N$ ) we use the result of Theorem 3:  $\sum_{j=0}^N p_{[j,j]}^* y_j = 1$ , which leads to:

$$y_0 = \frac{1}{p_{[0,0]}^* + \sum_{i=1, \rho_i \neq 1}^N \left[ \frac{(1 - \rho_i) \rho_i^n p_{[0,i]}^*}{(1 - \rho_i) + \rho_i (1 - \rho_i^n) p_{[i,i]}^*} \right] + \sum_{i=1, \rho_i = 1}^N \left[ \frac{p_{[0,i]}^*}{1 + n_i p_{[i,i]}^*} \right]} \quad (17)$$

#### Mean absorption time with absorbing barriers

We consider a star graph with starting point  $M[0]$  in the center and all other MFB's are absorbing:

$$\sum_{i=0}^N p_{[0,i]}^* = 1; \quad \prod_{i=1}^N p_{[0,i]}^* \neq 0; \quad p_{[i,i]}^* = 1 \quad (i = 1, 2, \dots, N)$$

We use notation:

$$p_i = p[0, i], \quad q_i = q[0, i], \quad n_i = n[0, i], \quad \rho_i = \frac{p_i}{q_i} \quad (i = 1, 2, \dots, N)$$

Let for  $i = 1, 2, \dots, N$ :

$$\alpha_i = \frac{1 - \rho_i^{n_i}}{1 - \rho_i^{n_i+1}}; \quad \beta_i = \frac{1 - (1 + n_i)\rho_i^{n_i} + n_i\rho_i^{n_i+1}}{(p_i - q_i)(1 - \rho_i^{n_i+1})} \quad (\rho_i \neq 1)$$

$$\alpha_i = \frac{n_i}{n_i + 1}; \quad \beta_i = -n_i \quad (\rho_i = 1)$$

Theorem 4 gives:

$$t_0 = \frac{\tau_0}{v_{00}} = \frac{1 - \sum_{i=1}^N \beta_i p_{[0,i]}^*}{1 - \sum_{i=1}^N \alpha_i p_{[0,i]}^*}$$

### 3.2 An infinite star graph with absorbing barriers

In this subsection we consider an infinite star graph where all barriers (except the start position  $M[0]$ ) are absorbing:  $p_{[i,i]}^* = 1 \quad (i = 1, 2, \dots)$ .

**Theorem 5 –**

$$y_0 = \frac{1}{1 - \sum_{i=1, \rho_i \neq 1}^{\infty} \left[ \frac{1 - \rho_i^n}{1 - \rho_i^{n+1}} \right] p_{[0,i]}^* - \sum_{i=1, \rho_i = 1}^{\infty} \left[ \frac{n}{1+n} \right] p_{[0,i]}^*} \quad (n = n_i) \quad (18)$$

$$y_i = \frac{(1 - \rho_i)\rho_i^n p_{[0,i]}^*}{1 - \rho_i^{n+1}} y_0 \quad (\rho_i \neq 1; \quad n = n_i; \quad i = 1, 2, \dots) \quad (19)$$

$$y_i = \frac{p_{[0,i]}^*}{1+n} y_0 \quad (\rho_i = 1; \quad n = n_i; \quad i = 1, 2, \dots) \quad (20)$$

*Proof.* Use (15),(16) and (17) with  $p_{[i,i]}^* = 1 (i = 1, 2, \dots, N)$  and rewrite (17) to the form in (18) by using  $p_{[i,i]}^* = 1$ . Finally note that  $\sum p_{[0,i]}^*$  is a majorant of both  $\sum \left[ \frac{n}{1+n} \right] p_{[0,i]}^*$  and  $\sum \left[ \frac{1 - \rho_i^n}{1 - \rho_i^{n+1}} \right] p_{[0,i]}^*$   $\square$

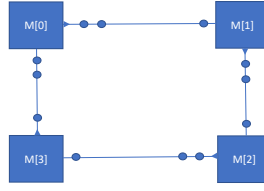
### 3. Examples of multiple function barrier graphs

#### 3.3 A positive oriented finite cycle graph

We have  $N + 1$  barriers in the finite cycle graph:  $M[0], M[1], \dots, M[N]$ . We start in  $M[0]$ . When the random walk is in  $M[i]$  then absorption can happen or we move one step in the direction of  $M[i + 1]$ . The state space is mod  $(N + 1)$ : when arriving in  $M[N]$ , there can be a step in the direction of  $M[N + 1]$  where  $M[N + 1] = M[0]$ . We have:

$$P_{[i,i]}^* + P_{[i,i+1]}^* = 1 \quad (i = 0, 1, \dots, N) \quad (21)$$

Example with  $N = 3$ :



We use the notation:  $\rho_i = \rho[i, i + 1]$ ;  $n_i = n[i, i + 1]$ .

#### Expected number of arrivals

**Theorem 6 –**

$$y_k = \frac{\prod_{i=1}^k M_i}{\sum_{m=0}^N P_{[m,m]}^* \prod_{i=1}^m M_i} \quad (k = 0, 1, \dots, N) \quad \left( \prod_{i=1}^0 M_i = 1 \right) \quad (22)$$

case  $\rho \neq 1$ :

$$M_i = \frac{P_{[i-1,i]}^*}{1 + \left[ \frac{1-\rho^n}{\rho^n(1-\rho)} \right] P_{[i,i]}^*} \quad (i = 1, 2, \dots, N) \quad \rho = \rho_i; \quad n = n_i \quad (23)$$

case  $\rho = 1$ :

$$M_i = \frac{P_{[i-1,i]}^*}{1 + n_i P_{[i,i]}^*} \quad (i = 1, 2, \dots, N) \quad (24)$$

*Proof.* We use (21) and Theorem 1 to obtain:  $y_i = -\frac{u_{i,i-1}}{u_{ii}} y_{i-1} \quad (i = 1, 2, \dots, N)$ . Let  $M_i = -\frac{u_{i,i-1}}{u_{ii}}$ ,  $\rho = \rho_i$  and  $n = n_i$  then  $M[i] = \frac{P_{[i-1,i]}^*}{1 + \left[ \frac{1-\rho^n}{(1-\rho)\rho^n} \right] P_{[i,i]}^*}$  ( $\rho \neq 1$ ) and  $M[i] = \frac{P_{[i-1,i]}^*}{1 + n_i P_{[i,i]}^*}$  ( $\rho = 1$ ). Because of  $y_i = M_i y_{i-1} \quad (i = 1, 2, \dots, N)$  we have  $y_k = \left( \prod_{i=1}^k M_i \right) y_0 \quad (k = 0, 1, \dots, N)$  where we define  $\prod_{i=1}^0 M_i = 1$ . By theorem 3 we have  $\sum_{j=0}^N P_{[j,j]}^* y_j = 1$ . Combining the last two results gives (22).  $\square$

### Mean absorption time

We use notation:

$$p_i = p[i, i+1], \quad q_i = q[i, i+1], \quad n_i = n[i, i+1], \quad \rho_i = \frac{p_i}{q_i} \quad (i = 0, 1, \dots, N)$$

Let for  $i = 0, 1, \dots, N$ :

$$\alpha_i = \frac{1 - \rho_i^{n_i}}{1 - \rho_i^{n_i+1}}; \quad \beta_i = \frac{1 - (1 + n_i)\rho_i^{n_i} + n_i\rho_i^{n_i+1}}{(p_i - q_i)(1 - \rho_i^{n_i+1})}; \quad \gamma_i = \frac{(1 - \rho_i)\rho_i^{n_i}}{1 - \rho_i^{n_i+1}} \quad (\rho_i \neq 1)$$

$$\alpha_i = \frac{n_i}{n_i + 1}; \quad \beta_i = -n_i; \quad \gamma_i = \frac{1}{n_i + 1} \quad (\rho_i = 1)$$

Using Theorem 4 we obtain:

$$v_{i,i+1} = \gamma_i p_{[i,i+1]}^*; \quad v_{i,i} = -1 + \alpha_i p_{[i,i+1]}^*; \quad \tau_i = -1 + \beta_i p_{[i,i+1]}^*$$

$$t_{i+1} = \frac{\tau_i - v_{i,i} t_i}{v_{i,i+1}} = \lambda_i t_i + \mu_i; \quad \lambda_i = \frac{-v_{i,i}}{v_{i,i+1}}; \quad \mu_i = \frac{\tau_i}{v_{i,i+1}} \quad (i = 0, 1, \dots, N)$$

$$t_{k+1} = t_0 \prod_{i=0}^k \lambda_i + \sum_{i=0}^{k-1} \mu_i \prod_{j=i+1}^k \lambda_j \quad (k = 0, 1, \dots, N)$$

$$t_0 = t_{N+1} = \frac{\sum_{i=0}^{N-1} \mu_i \prod_{j=i+1}^N \lambda_j}{1 - \prod_{i=0}^N \lambda_i}$$

### 3.4 A positive oriented infinite cycle graph

**Theorem 7** – CASE  $\rho \neq 1$ : If  $\sum_{i=0}^{\infty} p_{[i,i]}^*$  converges and  $\liminf_{i \rightarrow \infty} \rho_i > 0, \limsup_{i \rightarrow \infty, \rho_i < 1} \rho_i < 1, \liminf_{i \rightarrow \infty, \rho_i > 1} \rho_i > 1$ , then:

$$y_k = \frac{\prod_{i=1}^k M_i}{\sum_{m=0}^{\infty} p_{[m,m]}^* \prod_{i=1}^m M_i} \quad (k = 0, 1, \dots) \quad \left( \prod_{i=1}^0 M_i = 1 \right) \quad (25)$$

where  $M_i$  is given by (23).

CASE  $\rho = 1$ : We get (25) when  $\sum_{i=0}^{\infty} n_i p_{[i,i]}^*$  converges, where  $M_i$  is given by (24).

*Proof.* We use: If  $0 \leq \omega_i < 1$  then  $\prod_{i=1}^{\infty} (1 - \omega_i)$  converges to a non-zero number if and only if  $\sum_{i=1}^{\infty} \omega_i$  converges. We define a relation  $\sim$  between two sequences  $\sum a_i$

## References

and  $\sum b_i: \sum a_i \sim \sum b_i$  if and only if both sequences converges. Let  $\omega_i = 1 - M_i$ .  
CASE  $\rho \neq 1$ : Using 23 we get:

$$0 < \omega_i = \frac{(1 - \rho_i^n) p_{[i,i]}^* + (1 - \rho_i) \rho_i^n p_{[i-1,i-1]}^*}{(1 - \rho_i^n) p_{[i,i]}^* + (1 - \rho_i) \rho_i^n} < 1$$

If only a finite number of the  $\rho_i$  are in the neighbor of 0 and 1  
( $\liminf_{i \rightarrow \infty} \rho_i > 0, \limsup_{i \rightarrow \infty, \rho_i < 1} \rho_i < 1, \liminf_{i \rightarrow \infty, \rho_i > 1} \rho_i > 1$ ) then by the comparison criterium and  $\lim_{i \rightarrow \infty} p_{[i,i]}^* = 0$ :

$$\begin{aligned} \sum \omega_i &= \sum \frac{(1 - \rho_i^n) p_{[i,i]}^* + (1 - \rho_i) \rho_i^n p_{[i-1,i-1]}^*}{(1 - \rho_i^n) p_{[i,i]}^* + (1 - \rho_i) \rho_i^n} \sim \\ &\sum \frac{(1 - \rho_i^n) p_{[i,i]}^* + (1 - \rho_i) \rho_i^n p_{[i-1,i-1]}^*}{(1 - \rho_i) \rho_i^n} \sim \sum p_{[i,i]}^* \end{aligned}$$

CASE  $\rho = 1$ : Using 24 we get :  $0 < \omega_i = \frac{n_i p_{[i,i]}^* + p_{[i-1,i-1]}^*}{n_i p_{[i,i]}^* + 1} < 1$  By the comparison criterium and  $\lim_{i \rightarrow \infty} n_i p_{[i,i]}^* = 0$ :  $\sum \omega_i = \sum \frac{n_i p_{[i,i]}^* + p_{[i-1,i-1]}^*}{n_i p_{[i,i]}^* + 1} \sim \sum [n_i p_{[i,i]}^* + p_{[i-1,i-1]}^*] \sim \sum n_i p_{[i,i]}^*$  (because of  $n_i \geq 1$ ).  $\square$

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