

26. Computing the Rank of a Webpage

Google PageRank

More Practice with 2D Array OPs

More Practice with numpy

Functions and 2D Arrays

Assume

```
from random import uniform as randu  
from numpy import *
```

Let's write a function `randuM(m,n)` that returns an `m`-by-`n` array of random numbers, each chosen from the uniform distribution on `[0,1]`.

A Function that Returns an n-by-n Array of Random Numbers

```
def randuM(m,n):  
    A = zeros(m,n)  
    for i in range(m):  
        for j in range(n):  
            A[i,j] = randu(0,1)  
    return A
```

Probability Arrays

A $n \times n$ probability array has the property that its entries are nonnegative and that the sum of the entries in each column is 1

.2	.6	.2
.7	.3	.3
.1	.1	.5

Probability Arrays

To generate a random probability array, generate a random matrix with nonnegative entries and then divide the numbers in each column by the sum of the numbers in that column

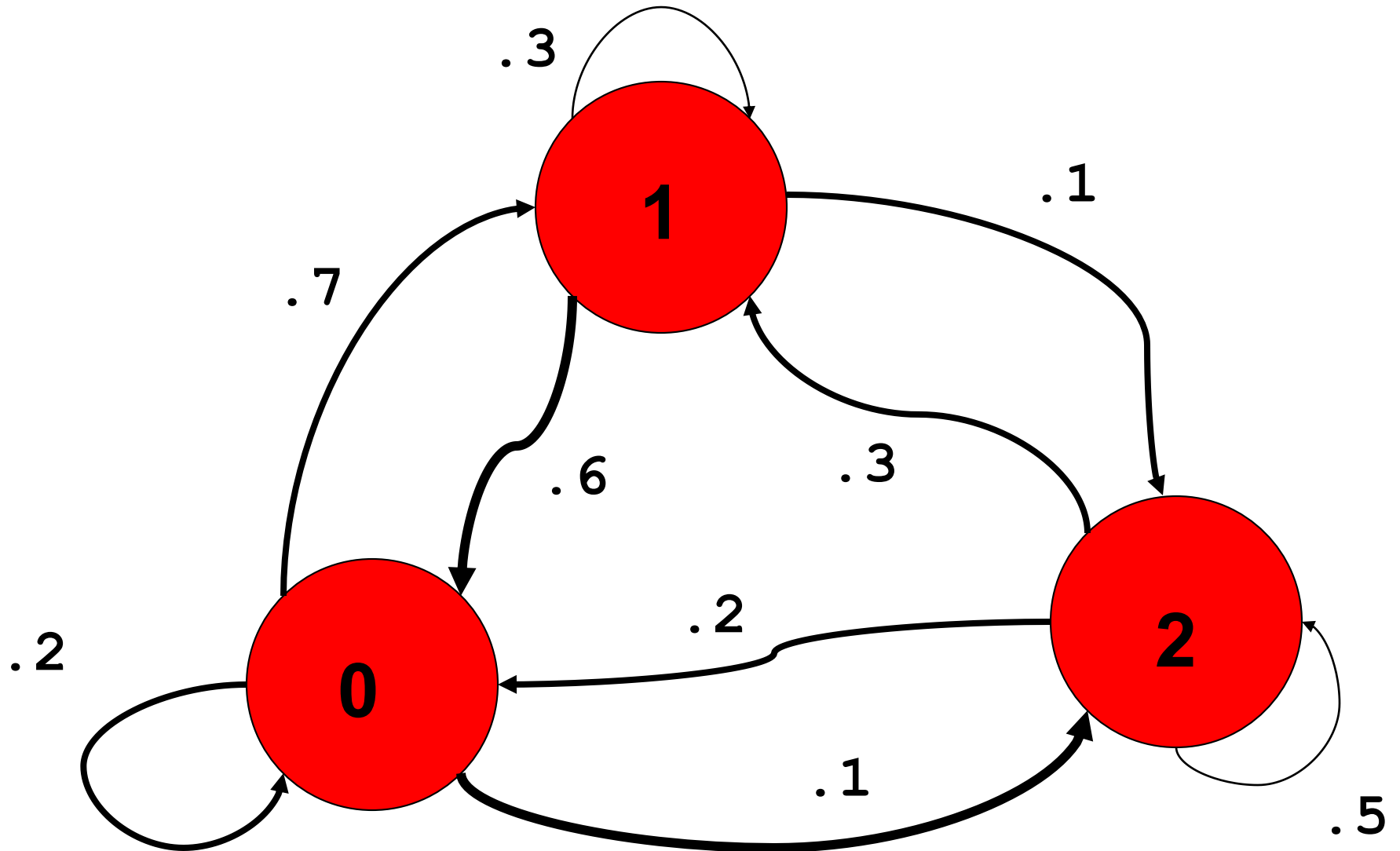
5	6	1
2	0	3
4	3	1

$5/11$	$6/9$	$1/5$
$2/11$	$0/9$	$3/5$
$4/11$	$3/9$	$1/5$

A Function that Returns a Random Probability Array

```
def probM(n) :  
    A = randuM(n,n)  
    for j in range(n) :  
        # Normalize column j  
        s = 0;  
        for i in range(n) :  
            s += A[i,j]  
        for i in range(n) :  
            A[i,j] = A[i,j]/s  
    return A
```

Here is a Network



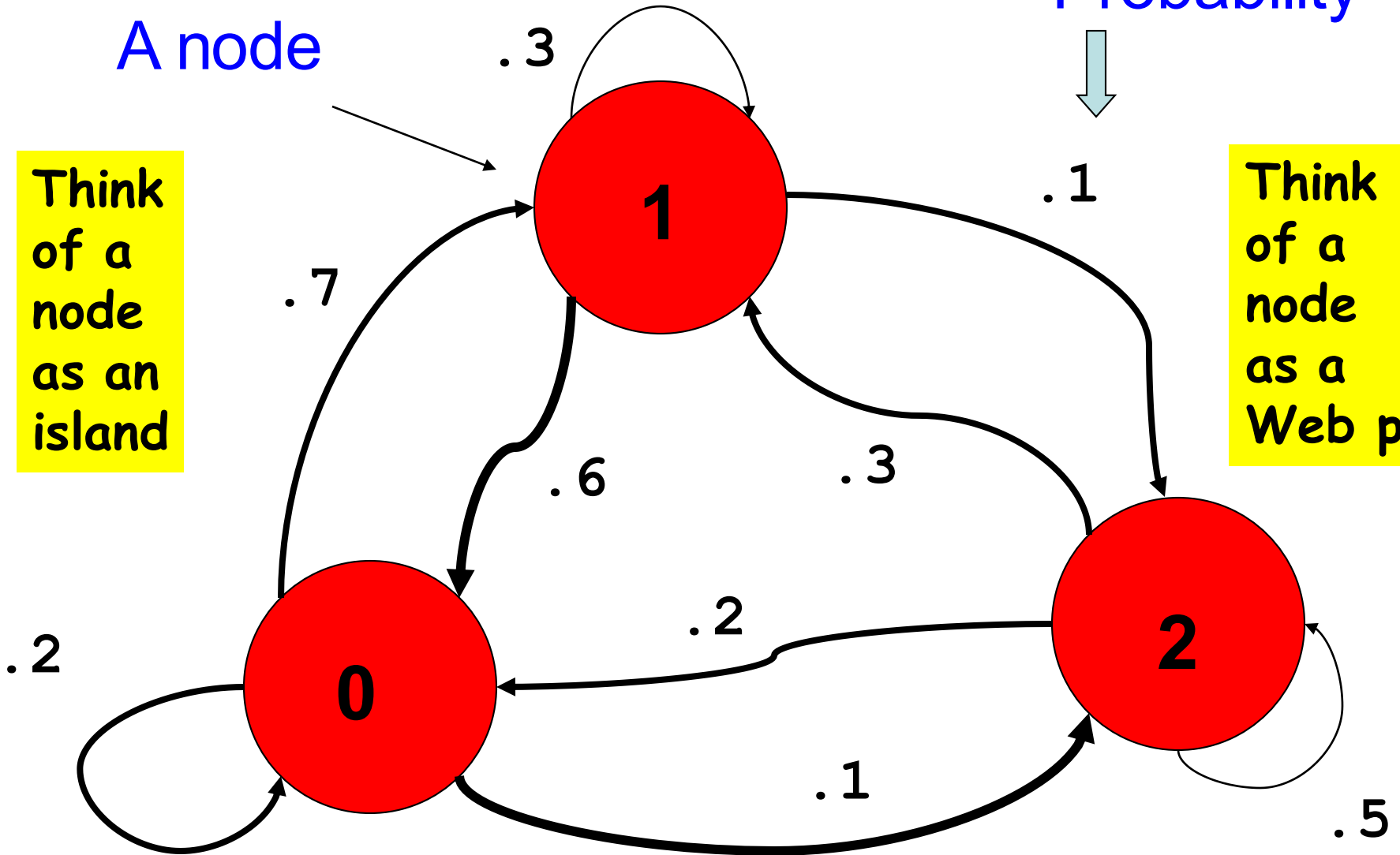
A Transition Probability



A node

Think of a node as an island

Think of a node as a Web page

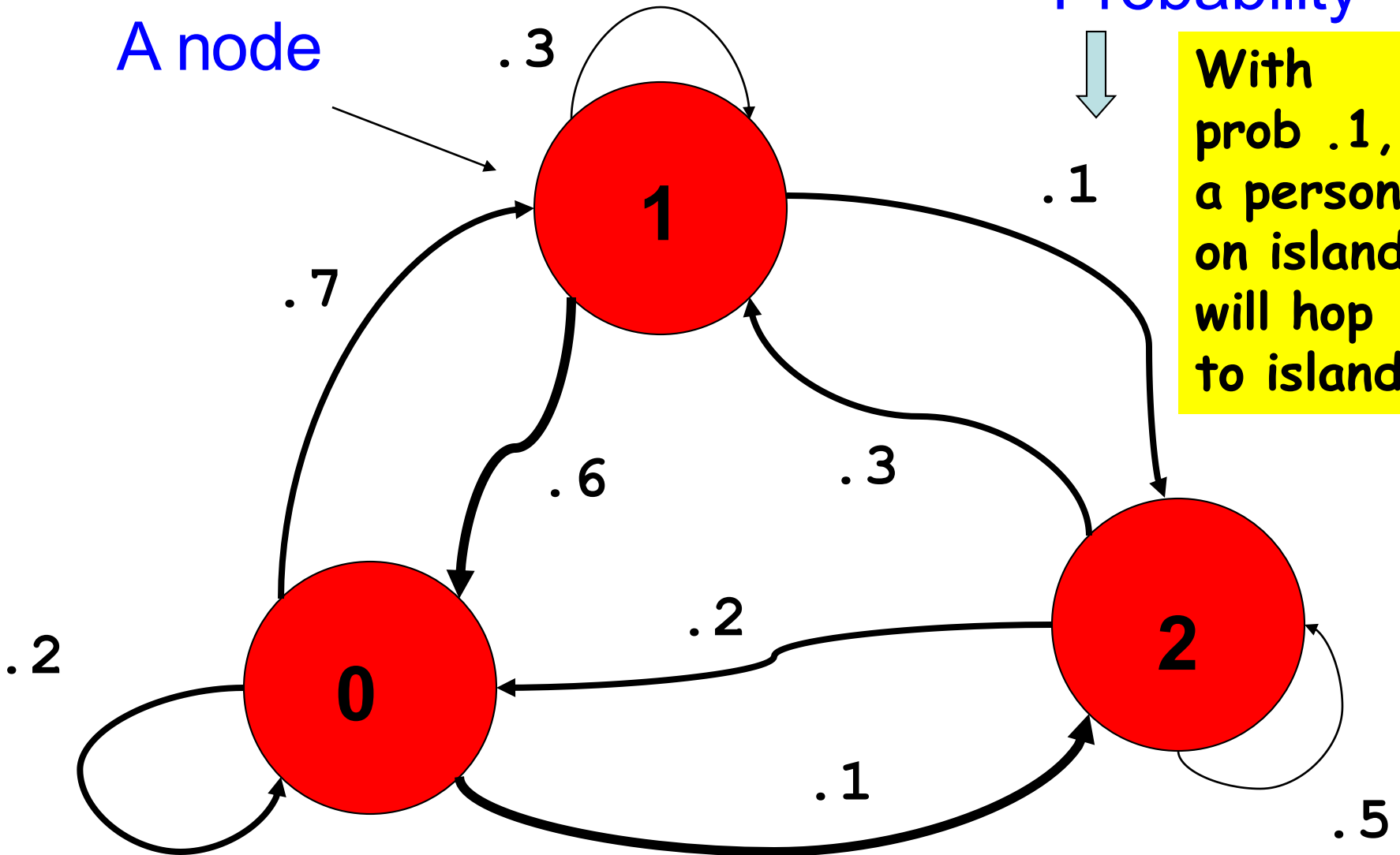


A Transition Probability



With prob .1, a person on island 1 will hop to island 2

A node

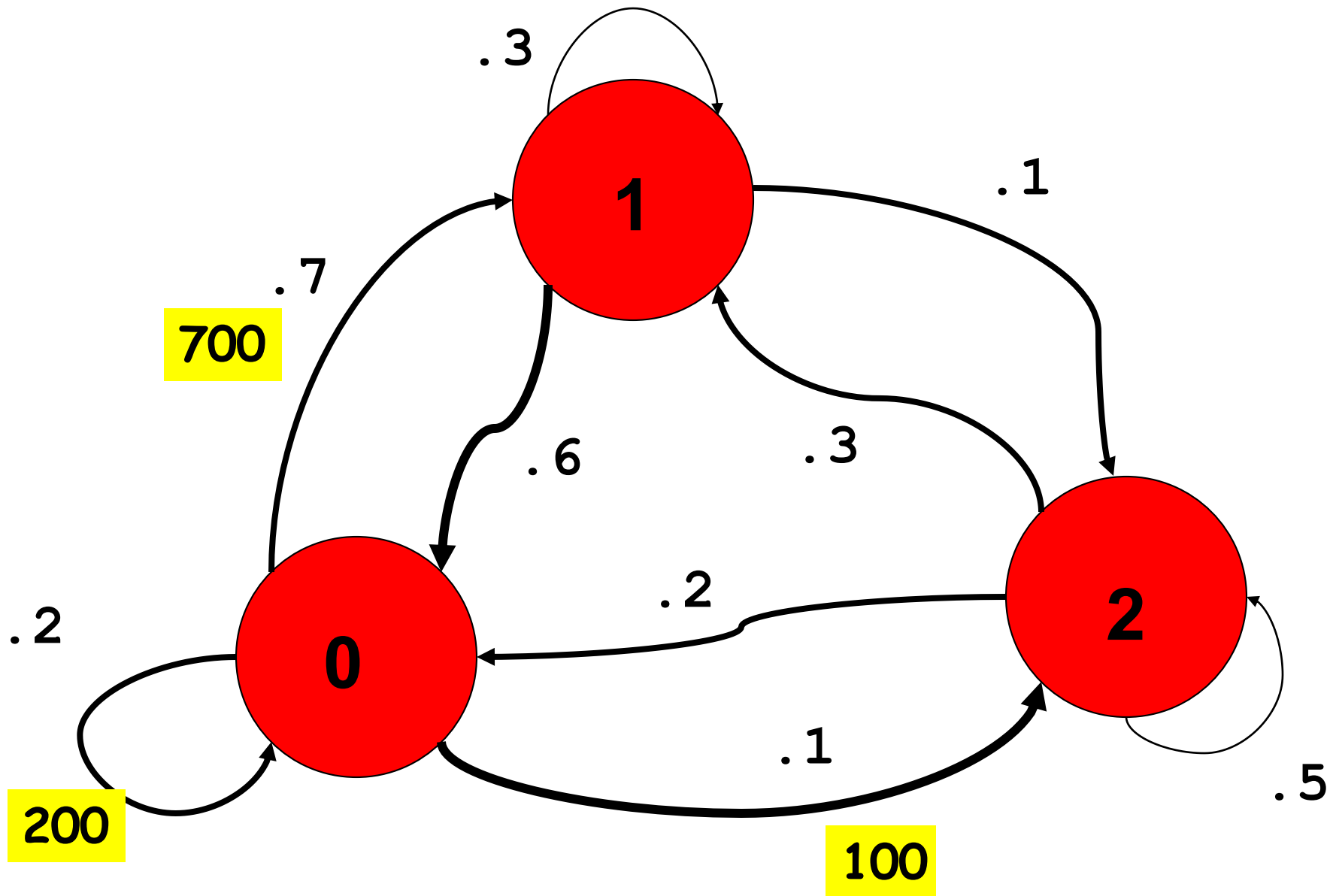


A Random Process

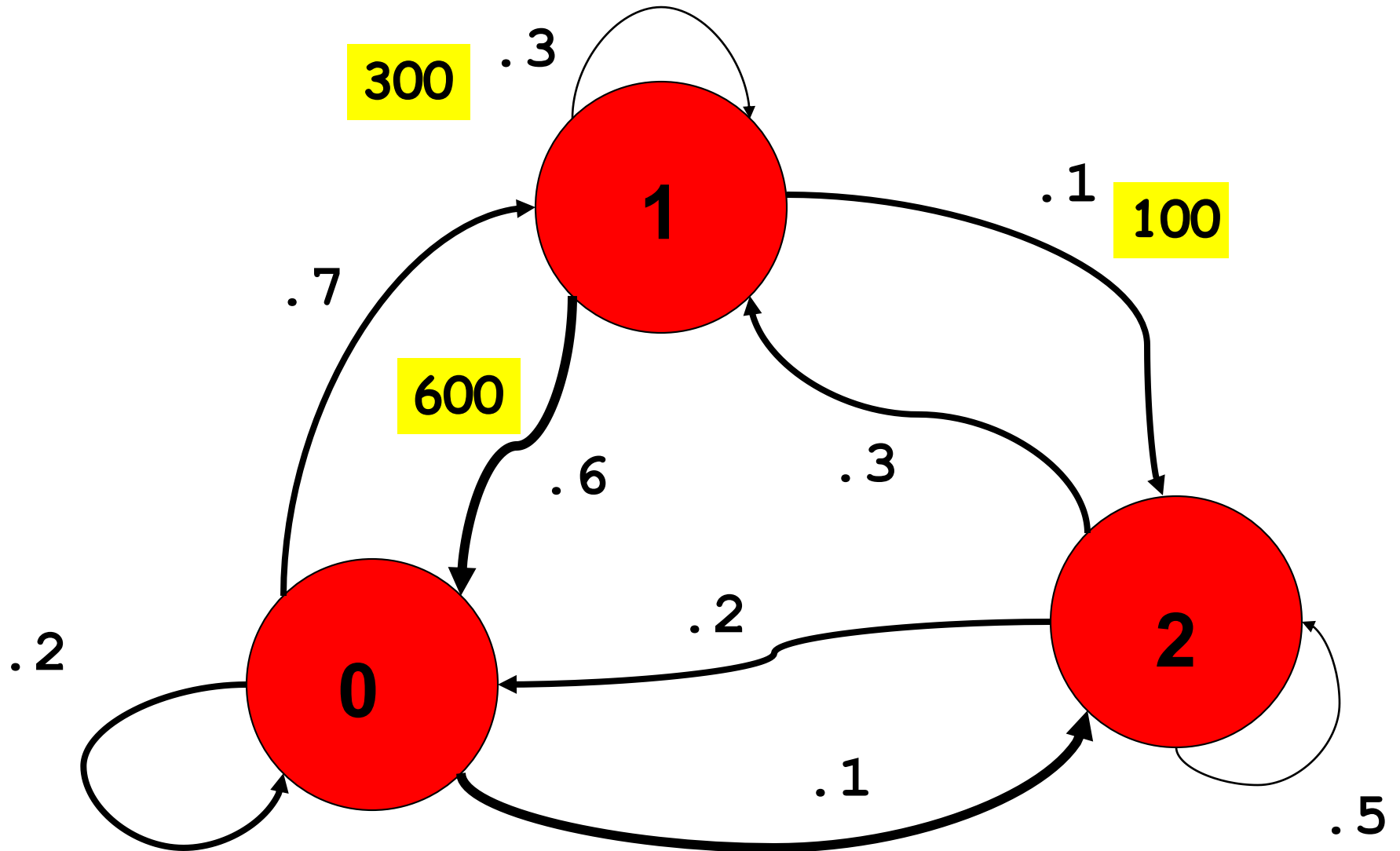
Suppose there are a 1000 people on each node.

At the sound of a whistle they hop to another node in accordance with the “outbound” probabilities.

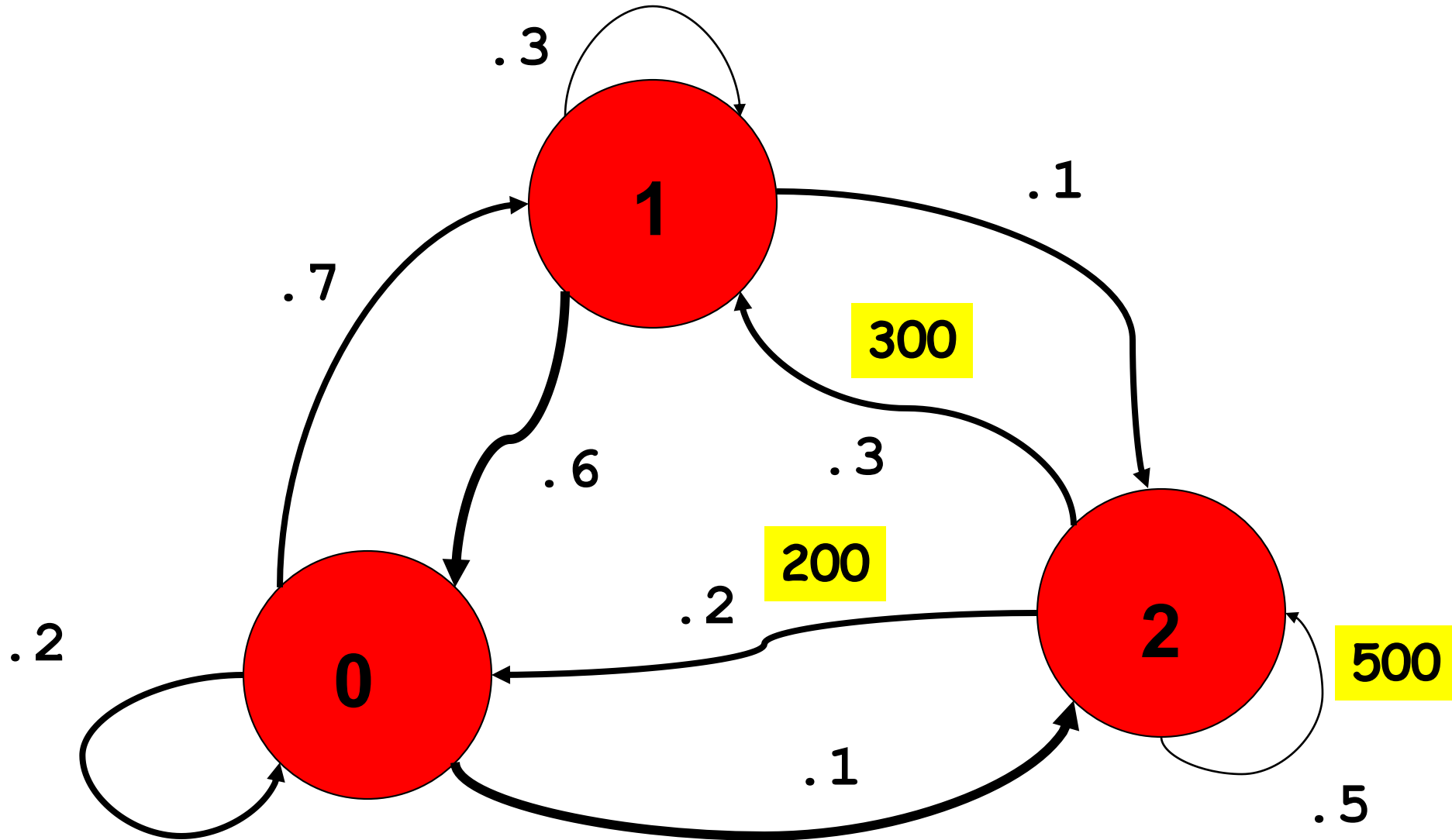
At Node 0



At Node 1



At Node 2



The Population Distribution

	Before	After
Node 0	1000	1000
Node 1	1000	1300
Node 2	1000	700

Repeat

	Before	After
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Node 0	1000	1120
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Node 1	1300	1300
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Node 2	700	580
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After 100 Iterations

	Before	After
Node 0	1142.85	1142.85
Node 1	1357.14	1357.14
Node 2	500.00	500.00

Appears to reach a Steady State

After 100 Iterations

Before

After

Node 0 1142.85 1142.85

Node 1 1357.14 1357.14

Node 2 500.00 500.00

In terms of popularity: Island 1 > Island 0 > Island 2

After 100 Iterations

Before

After

Node 0 1142.85 1142.85

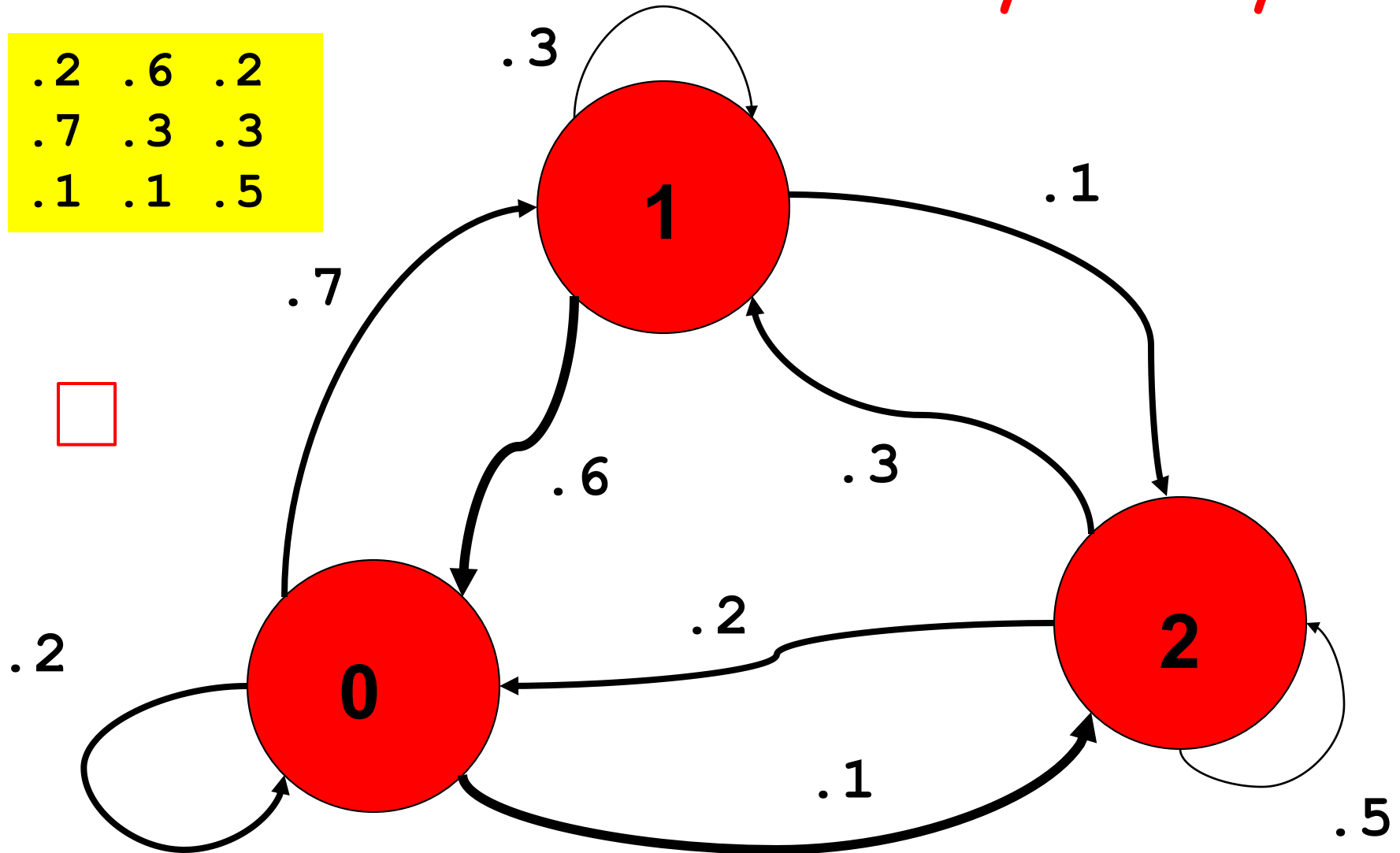
Node 1 1357.14 1357.14

Node 2 500.00 500.00

[1142.85, 1357.14, 500.0] is the "stationary array"

Computing the Stationary Array Involves a Probability Array

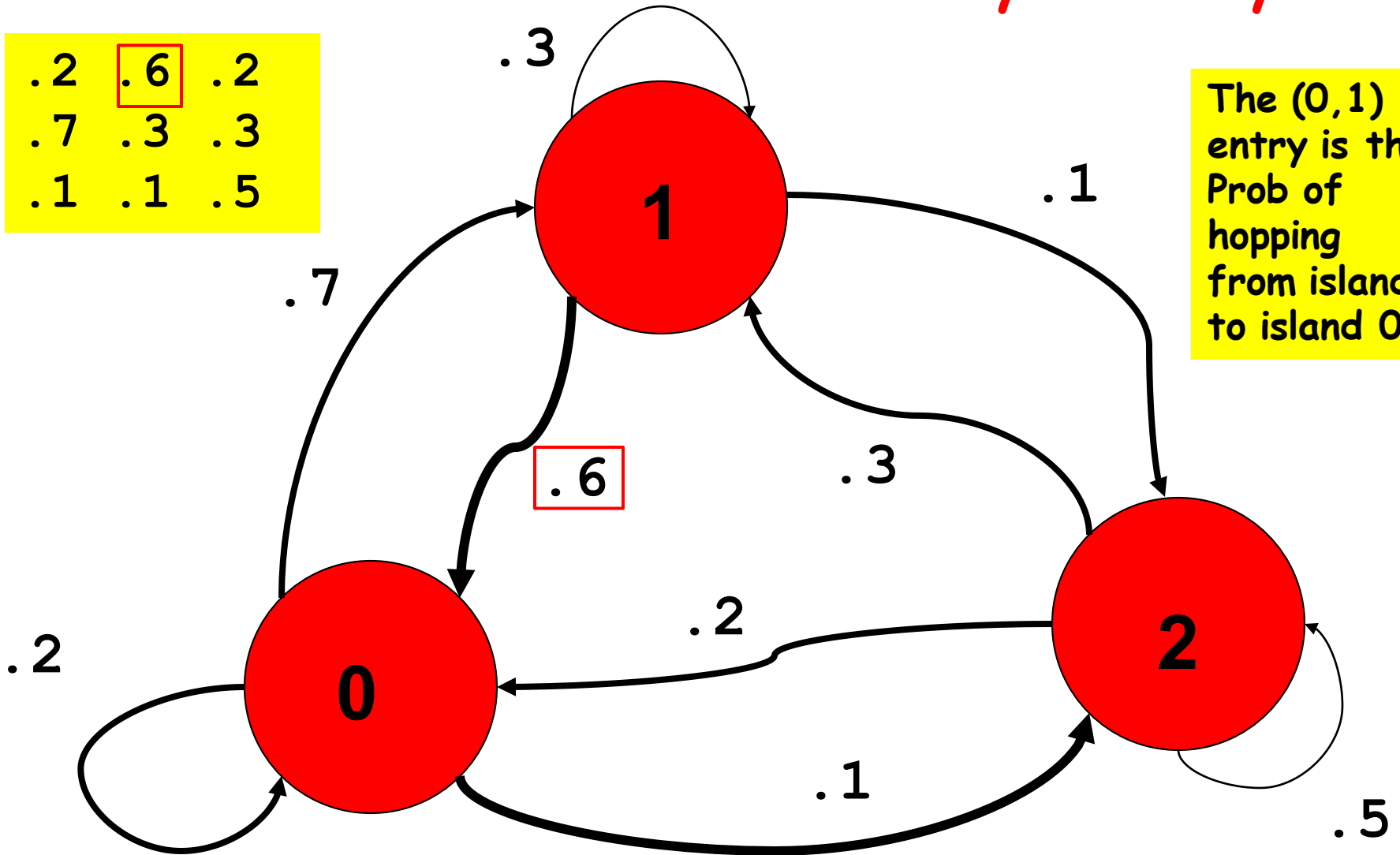
.2	.6	.2
.7	.3	.3
.1	.1	.5



Computing the Stationary Array Involves a Probability Array

.2	.6	.2
.7	.3	.3
.1	.1	.5

The (0,1) entry is the Prob of hopping from island 1 to island 0



Transition Probability Array

P:

.2	.6	.2
.7	.3	.3
.1	.1	.5

$P[i,j]$ is the probability of hopping from node j to node i

Formula for Updating the Distribution Array

$$P = \begin{array}{|c|c|c|} \hline .2 & .6 & .2 \\ \hline .7 & .3 & .3 \\ \hline .1 & .1 & .5 \\ \hline \end{array}$$

$$w[0] = .2*v[0] + .6*v[1] + .2*v[2]$$

$$w[1] = .7*v[0] + .3*v[1] + .3*v[2]$$

$$w[2] = .1*v[0] + .1*v[1] + .5*v[2]$$

**V is the old distribution array,
w is the updated distribution array**

Formula for Updating the Distribution Vector

$$P = \begin{array}{|c|c|c|} \hline .2 & .6 & .2 \\ \hline .7 & .3 & .3 \\ \hline .1 & .1 & .5 \\ \hline \end{array}$$

$$w[0] = P[0,0]*v[0] + P[0,1]*v[1] + P[0,2]*v[2]$$

$$w[1] = P[1,0]*v[0] + P[1,1]*v[1] + P[1,2]*v[2]$$

$$w[2] = P[2,0]*v[0] + P[2,1]*v[1] + P[2,2]*v[2]$$

**V is the old distribution vector,
w is the updated distribution vector**

A Function that Computes the Update

```
def Update (P, v) :  
    n = len(x)  
    w = zeros ( (n, 1) )  
    for i in range(n) :  
        for j in range(n) :  
            w[i] += P[i, j]*v[j]  
    return w
```


Back to PageRank

Background

Index all the pages on the Web from 0 to $N-1$. (N is around 50 billion.)

The PageRank algorithm orders these pages from "most important" to "least important".

It does this by analyzing links, not content.

Key Ideas

The Transition Probability Array

A Very Special Random Walk

The Connectivity Array

A Random Walk on the Web

Repeat:

You are on a webpage.

There are m outlinks.

Choose one at random.

Click on the link.

The Connectivity Array

$G[i,j]$ is
1 if there
is a link
on page j
to page i

G:

0	1	0	0	1	0	1	0
1	0	0	0	0	0	1	1
0	1	0	0	1	0	0	0
1	0	1	1	0	1	0	0
0	0	0	1	0	0	1	0
0	1	1	0	0	1	0	0
1	0	0	0	0	0	1	0
0	0	1	0	0	1	0	0

The Probability Array

$$a = 1/3$$

$$b = 1/2$$

$$c = 1/4$$

P:

0	a	0	0	b	0	c	0
a	0	0	0	0	0	c	1
0	a	0	0	b	0	0	0
a	0	a	b	0	a	0	0
0	0	0	b	0	0	c	0
0	a	a	0	0	a	0	0
a	0	0	0	0	0	c	0
0	0	a	0	0	a	0	0

PageRank From the Stationary Array

