

# Link Analysis

Introduction to Information Retrieval  
CS 221  
Donald J. Patterson

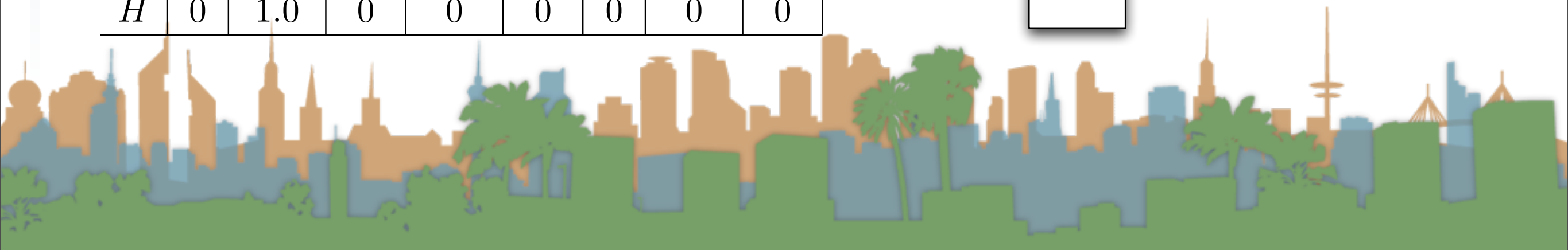
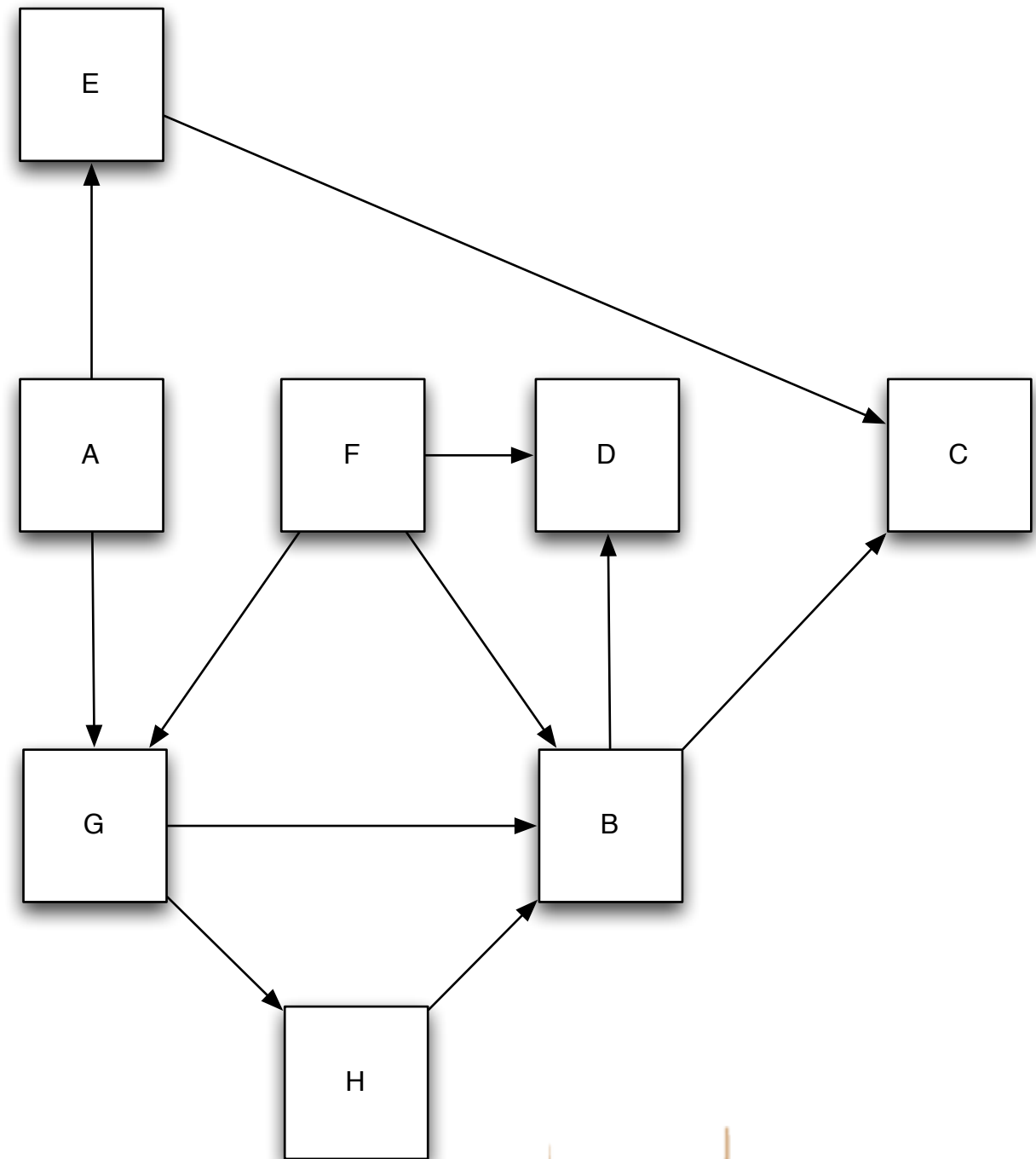
Content adapted from Hinrich Schütze  
<http://www.informationretrieval.org>



## Markov Chains

- Example:
  - 8 states
    - (web pages or whatever)
  - 8 by 8 transition prob. matrix

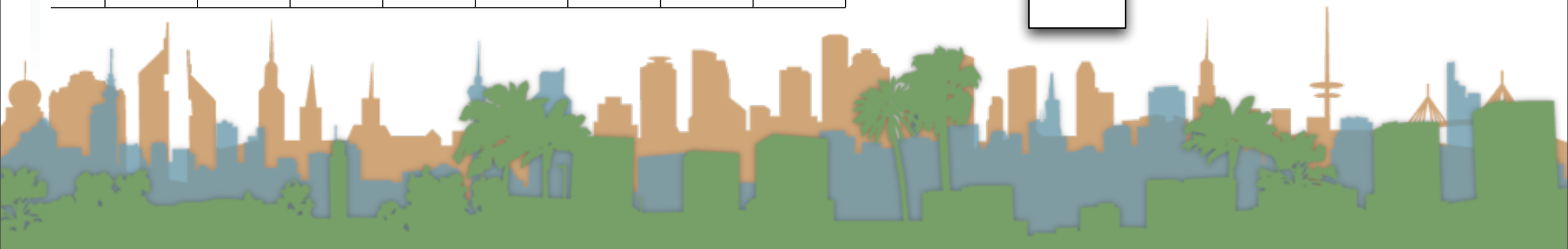
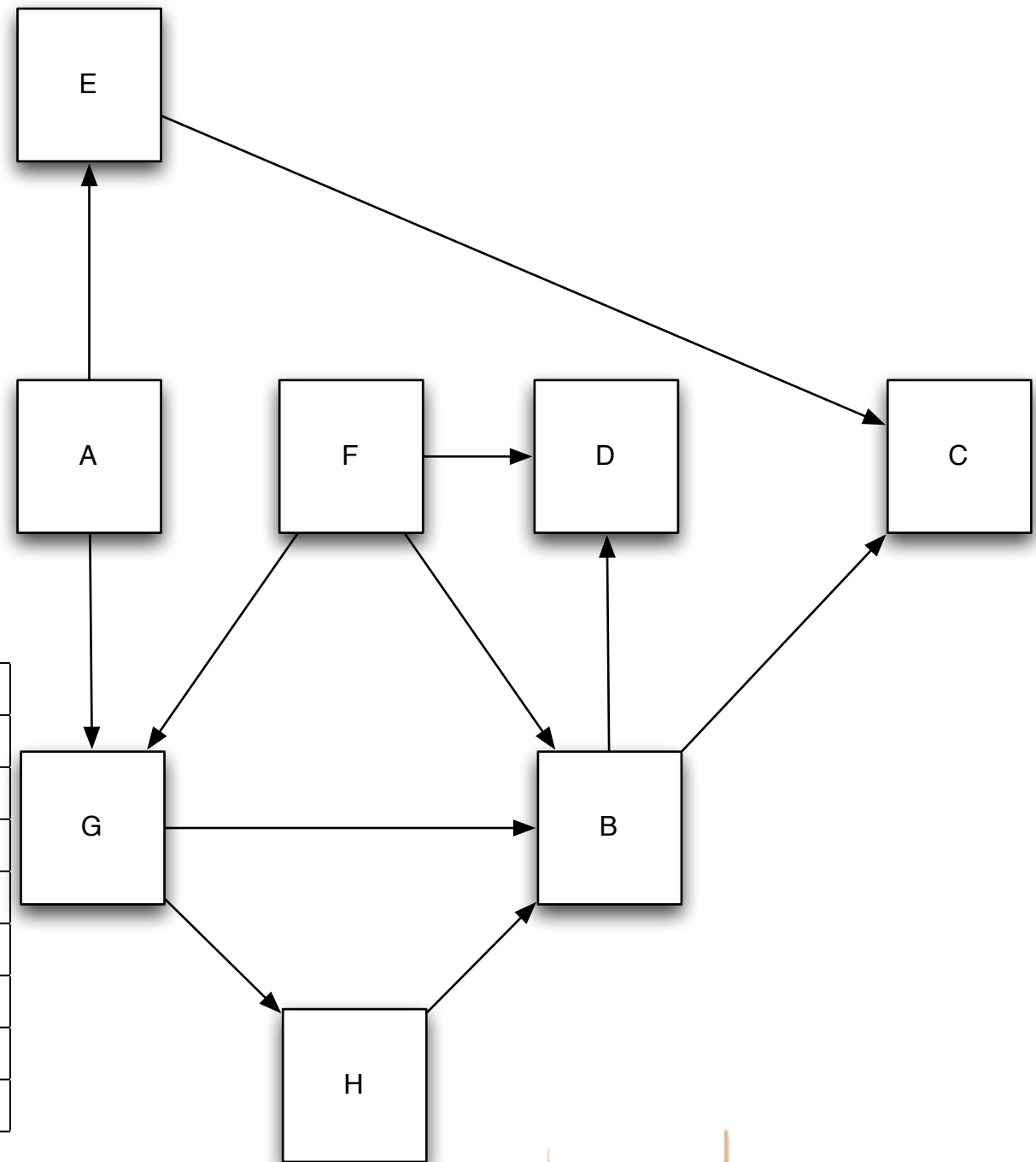
	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0	0	0	0	0	0	0	0
<i>D</i>	0	0	0	0	0	0	0	0
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0



## Markov Chains

- Example:
  - 8 states
  - 8 by 8 transition prob. matrix
  - Handle Dead-Ends also

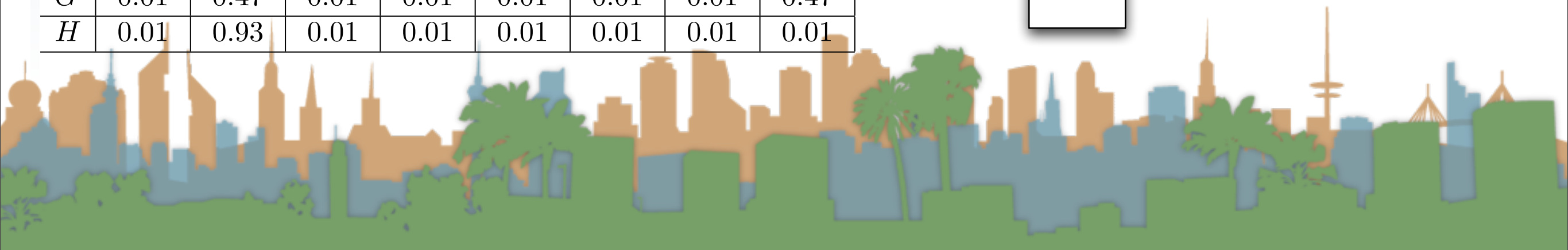
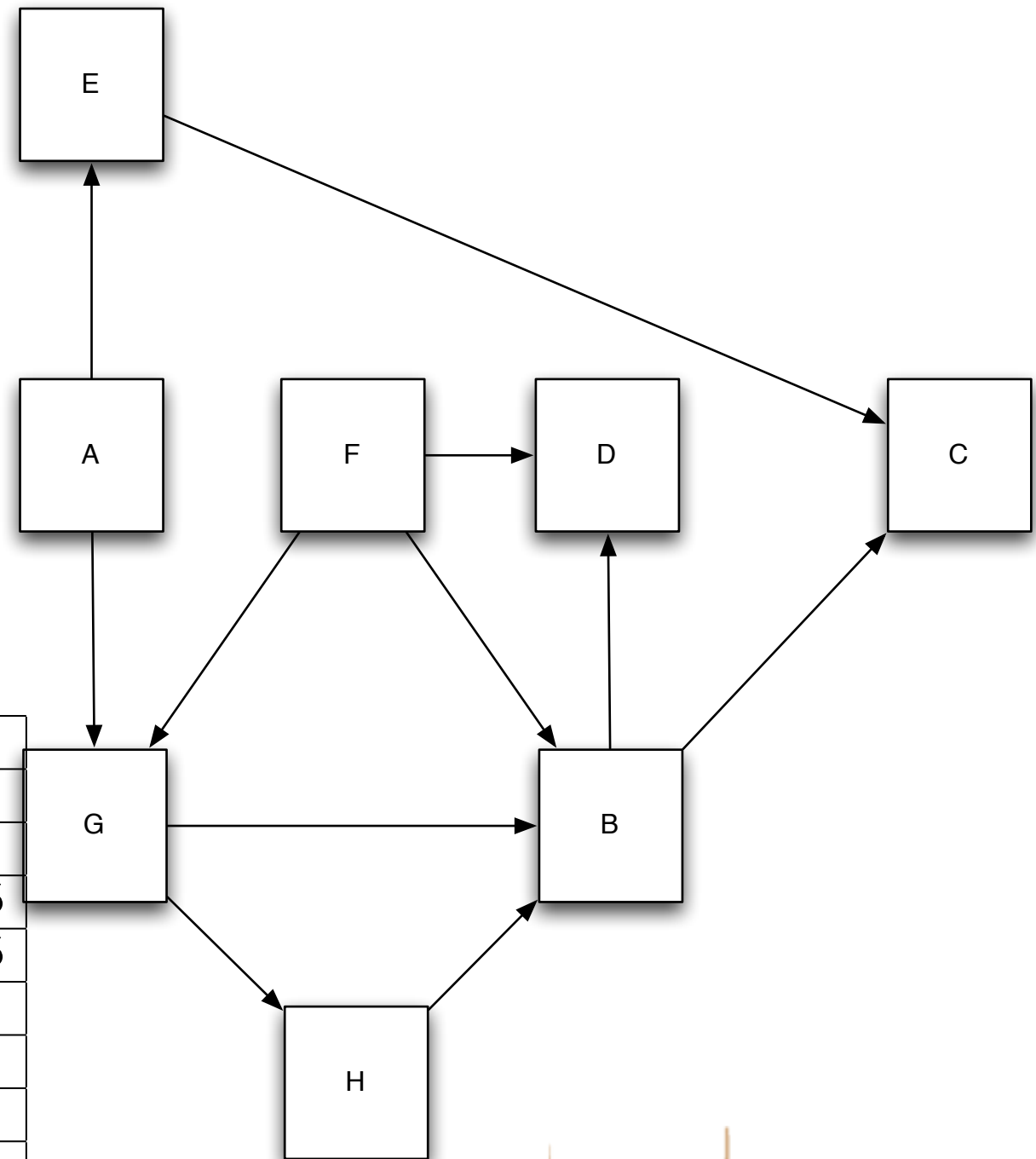
	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0



## Markov Chains

- Example:
  - 8 states
  - 8 by 8 transition prob. matrix
  - Handle Dead-Ends also
  - Handle teleports

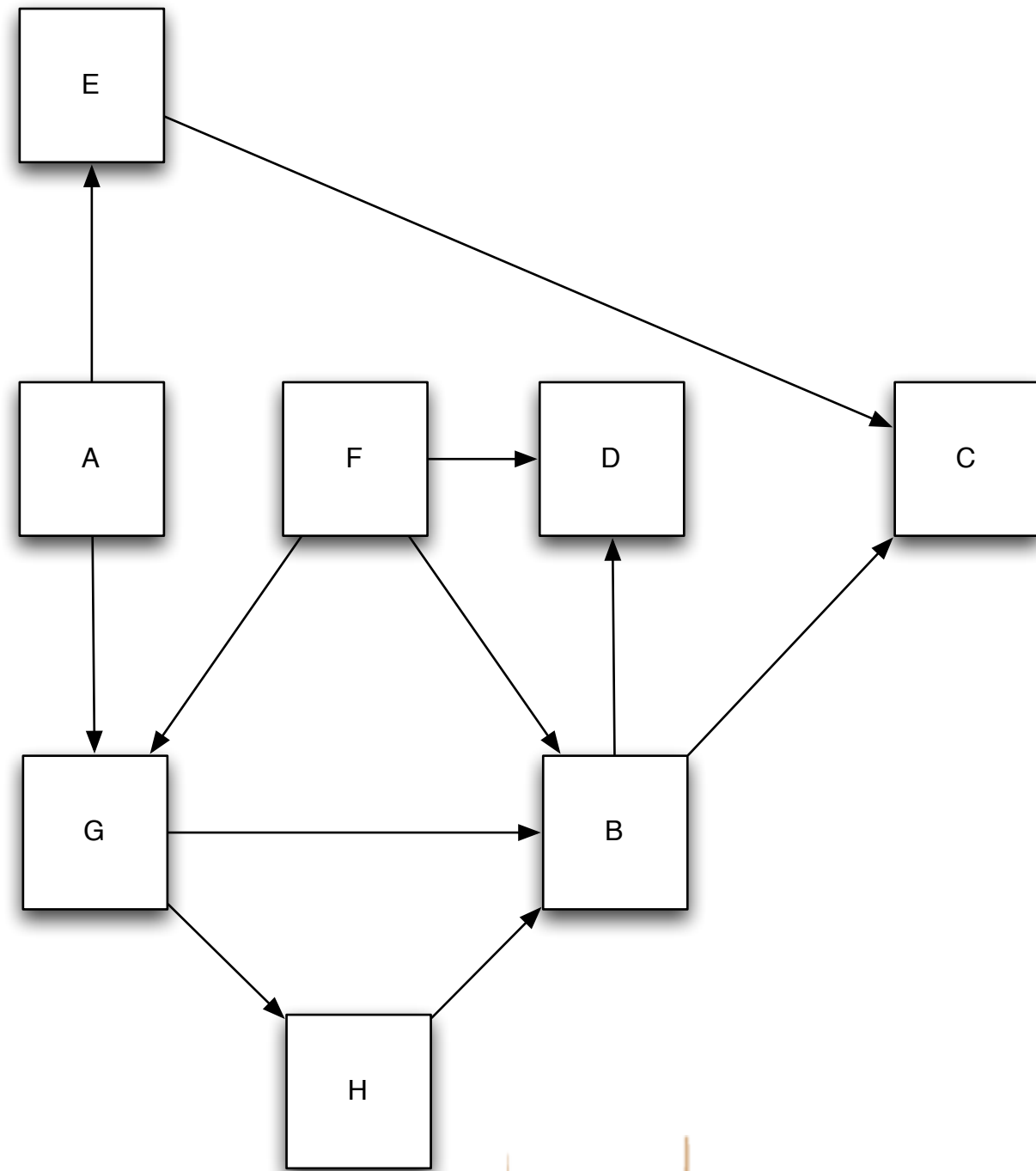
	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0.01	0.01	0.01	0.01	0.47	0.01	0.47	0.01
<i>B</i>	0.01	0.01	0.47	0.47	0.01	0.01	0.01	0.01
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0.01	0.01	0.93	0.01	0.01	0.01	0.01	0.01
<i>F</i>	0.01	0.32	0.01	0.32	0.01	0.01	0.32	0.01
<i>G</i>	0.01	0.47	0.01	0.01	0.01	0.01	0.01	0.47
<i>H</i>	0.01	0.93	0.01	0.01	0.01	0.01	0.01	0.01



## Markov Chain : The Game

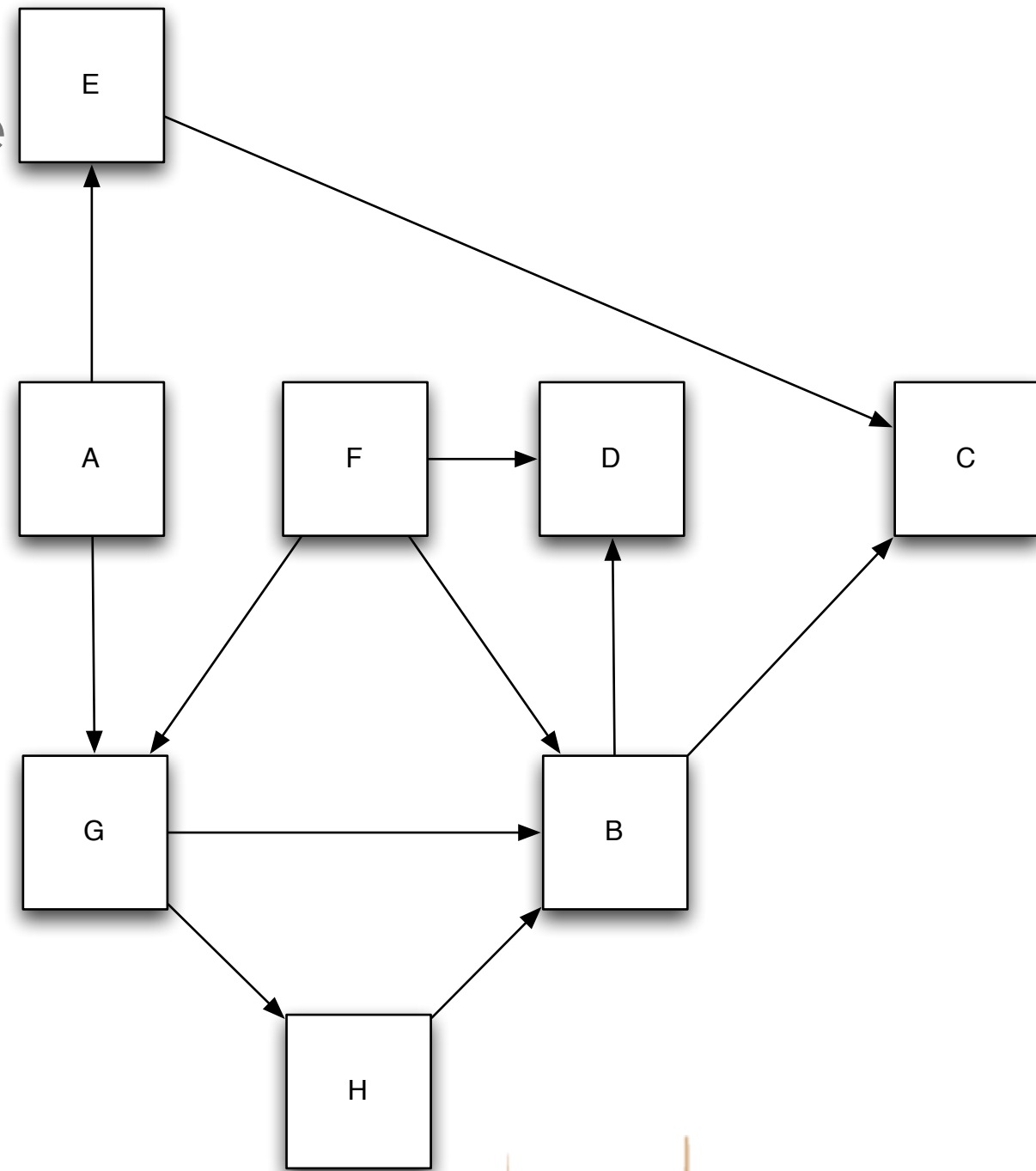
- You may be in one state at a time
- Every tick you move one step  
chosen randomly from the  
transition probability matrix

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0



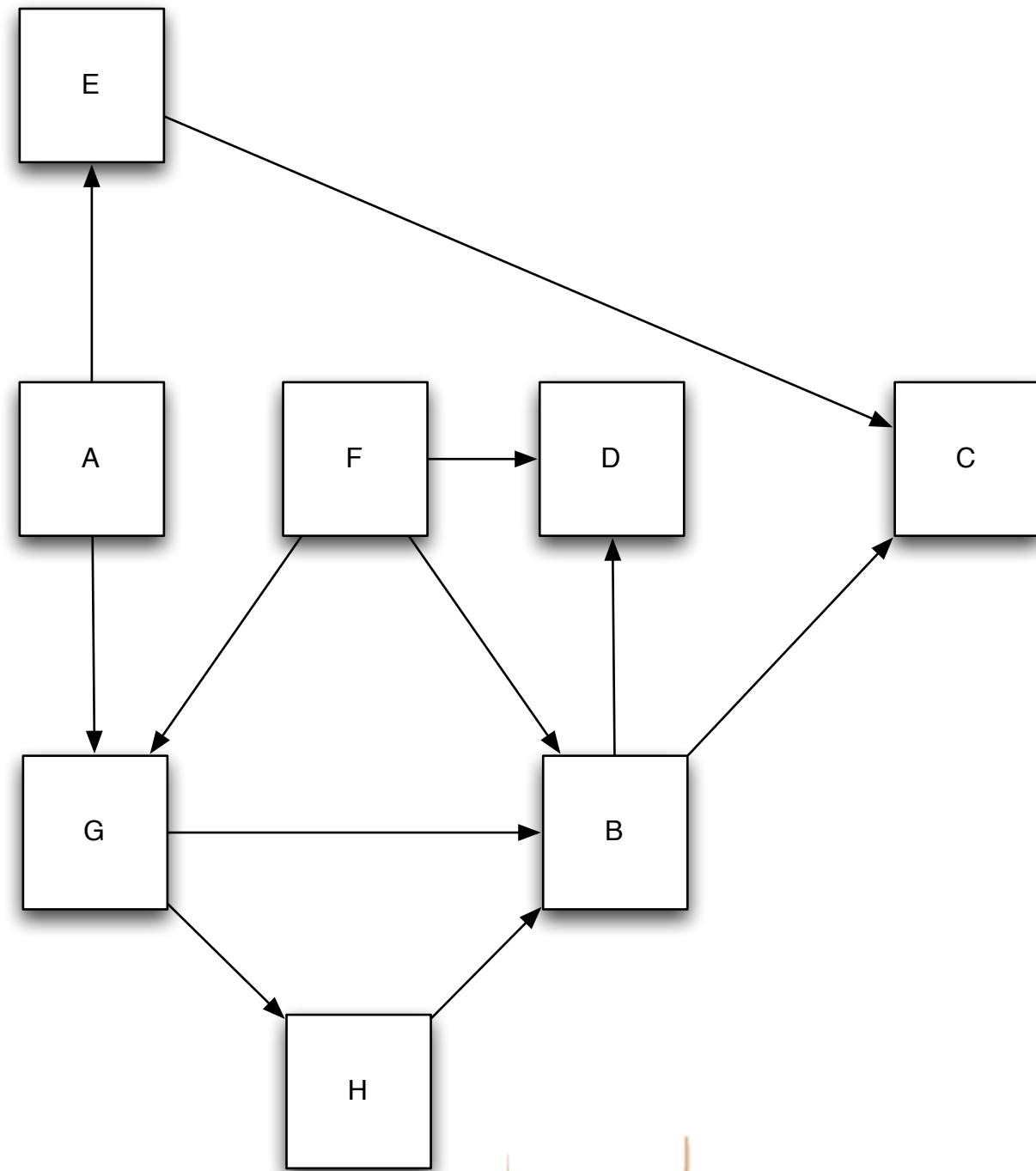
## The Markov Property

- It doesn't matter where you came from.
- All information that you need to take the next step comes from your current state and the transition probability matrix
- History is irrelevant given your current state

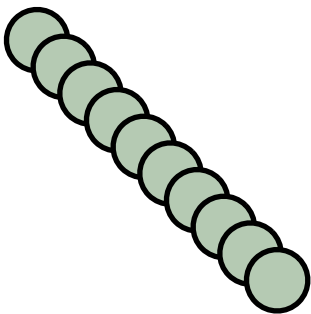


## PageRank

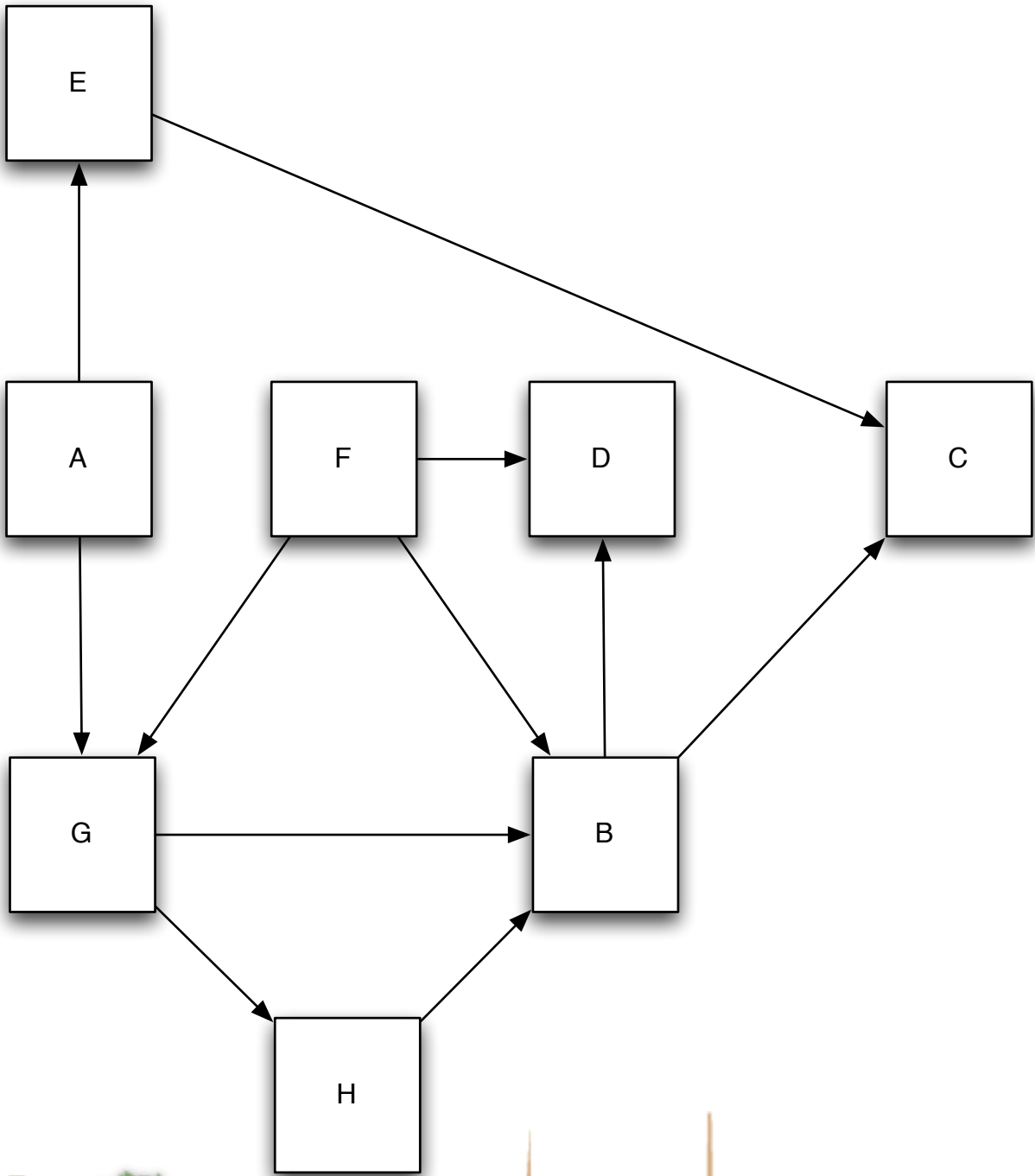
- PageRank is the long term visit rate of a random walk on the graph.
- With teleports



# Long-Term visit rate



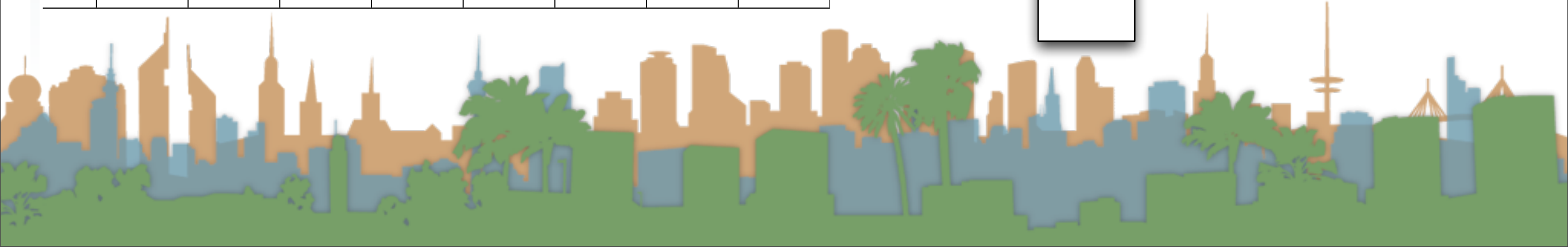
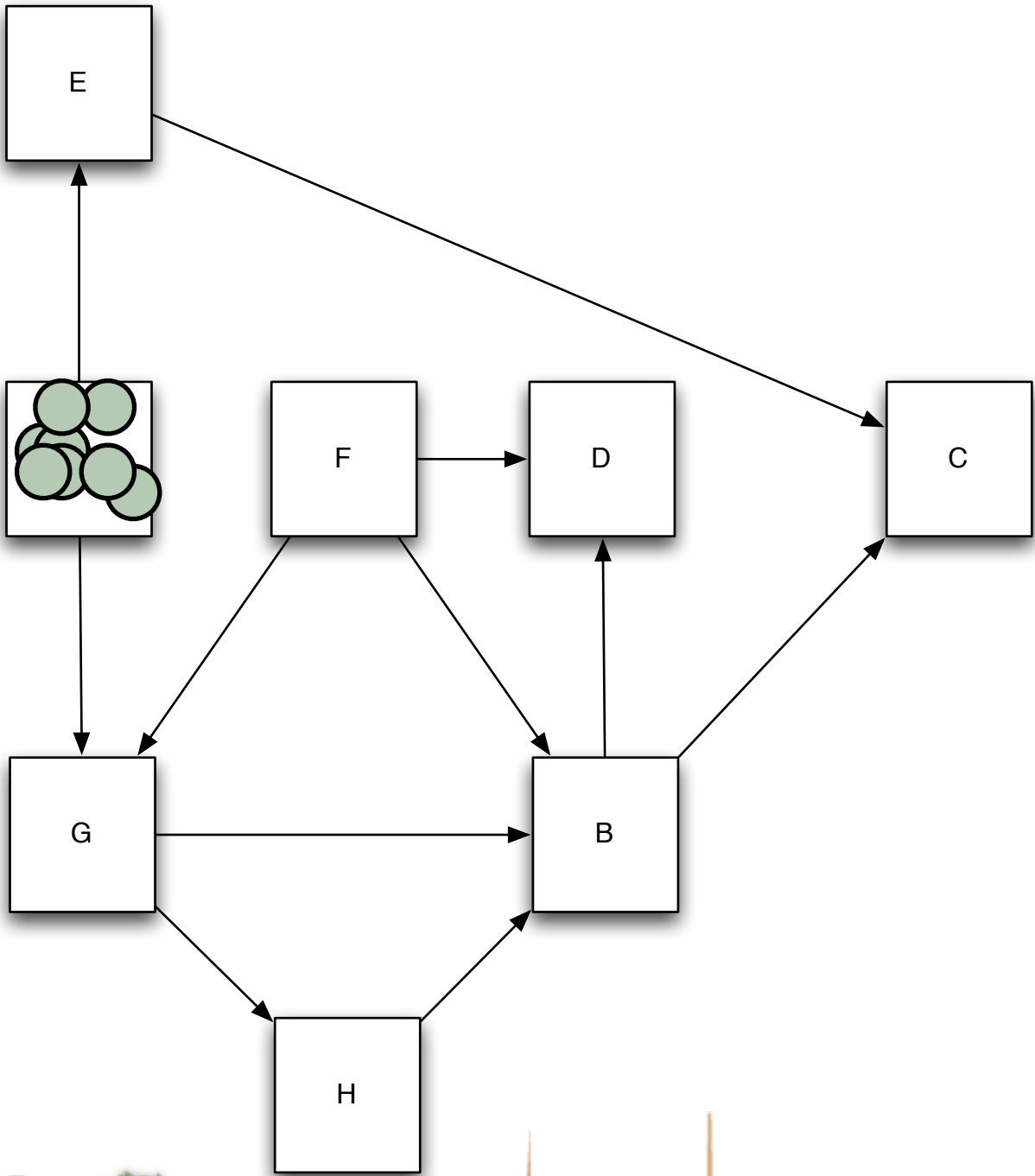
	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0





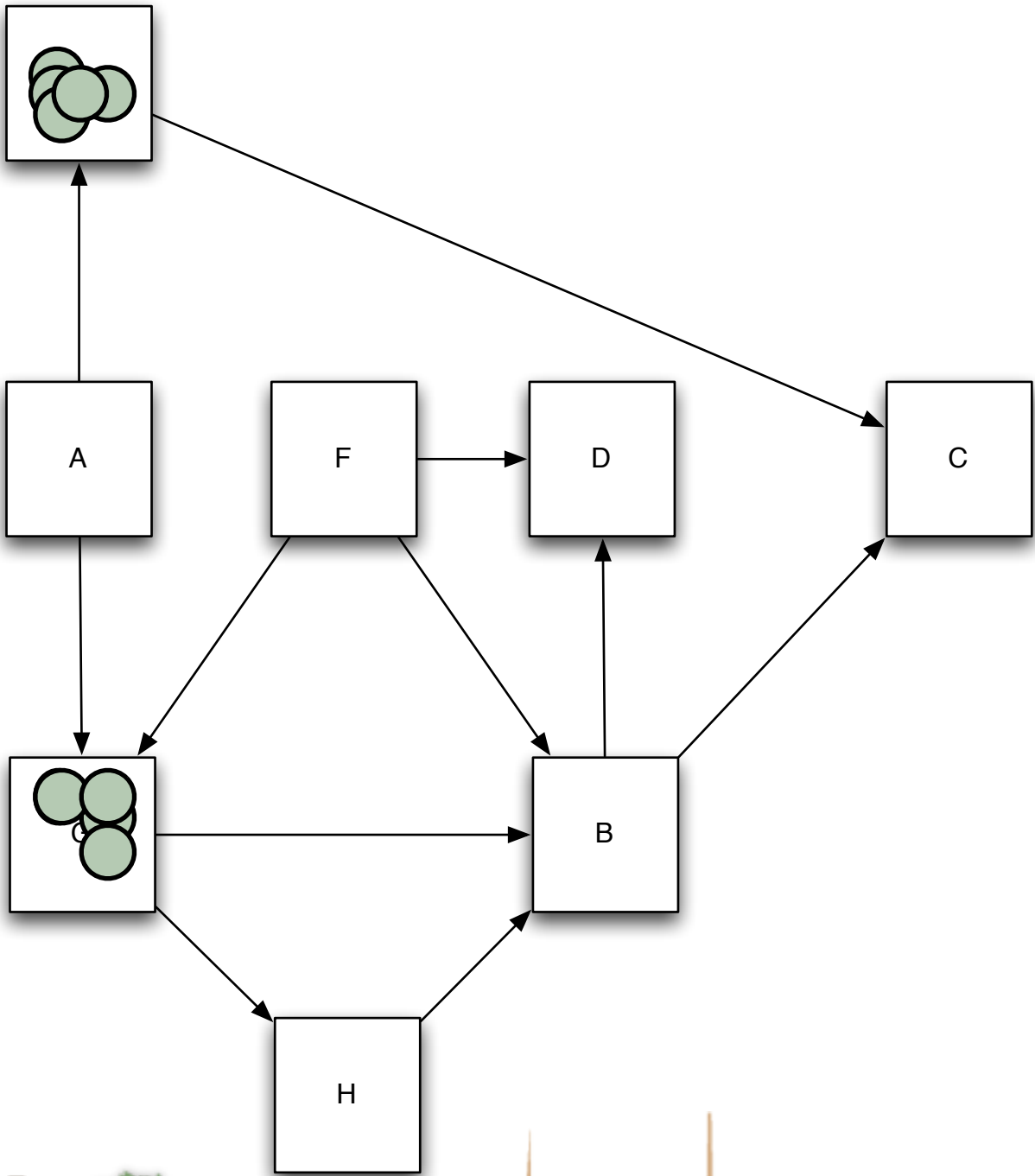
Long-Term visit rate

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0



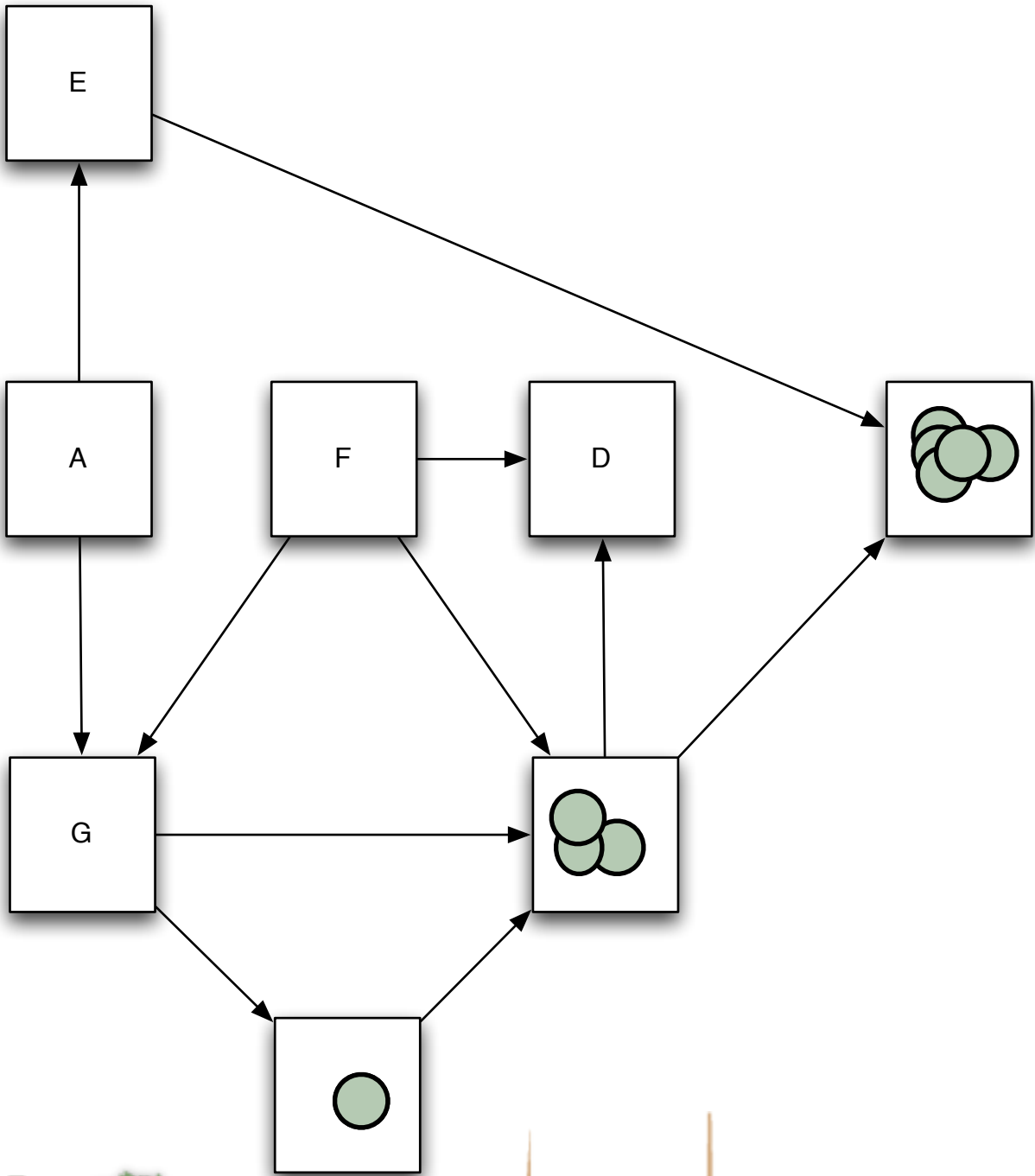
# Long-Term visit rate

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0



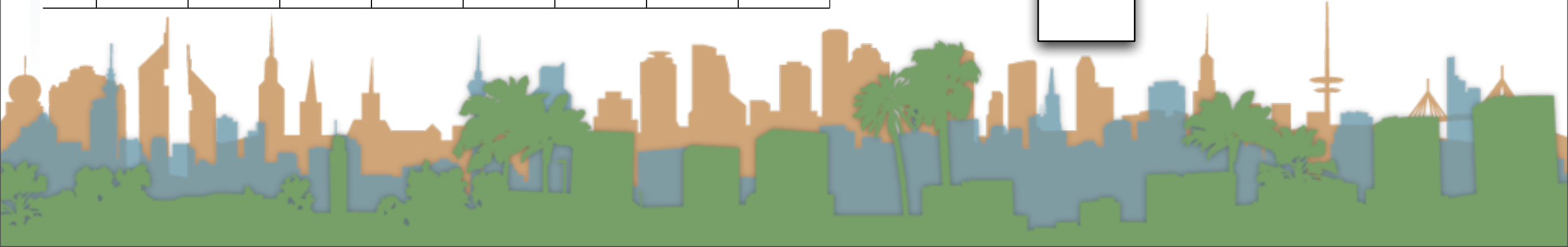
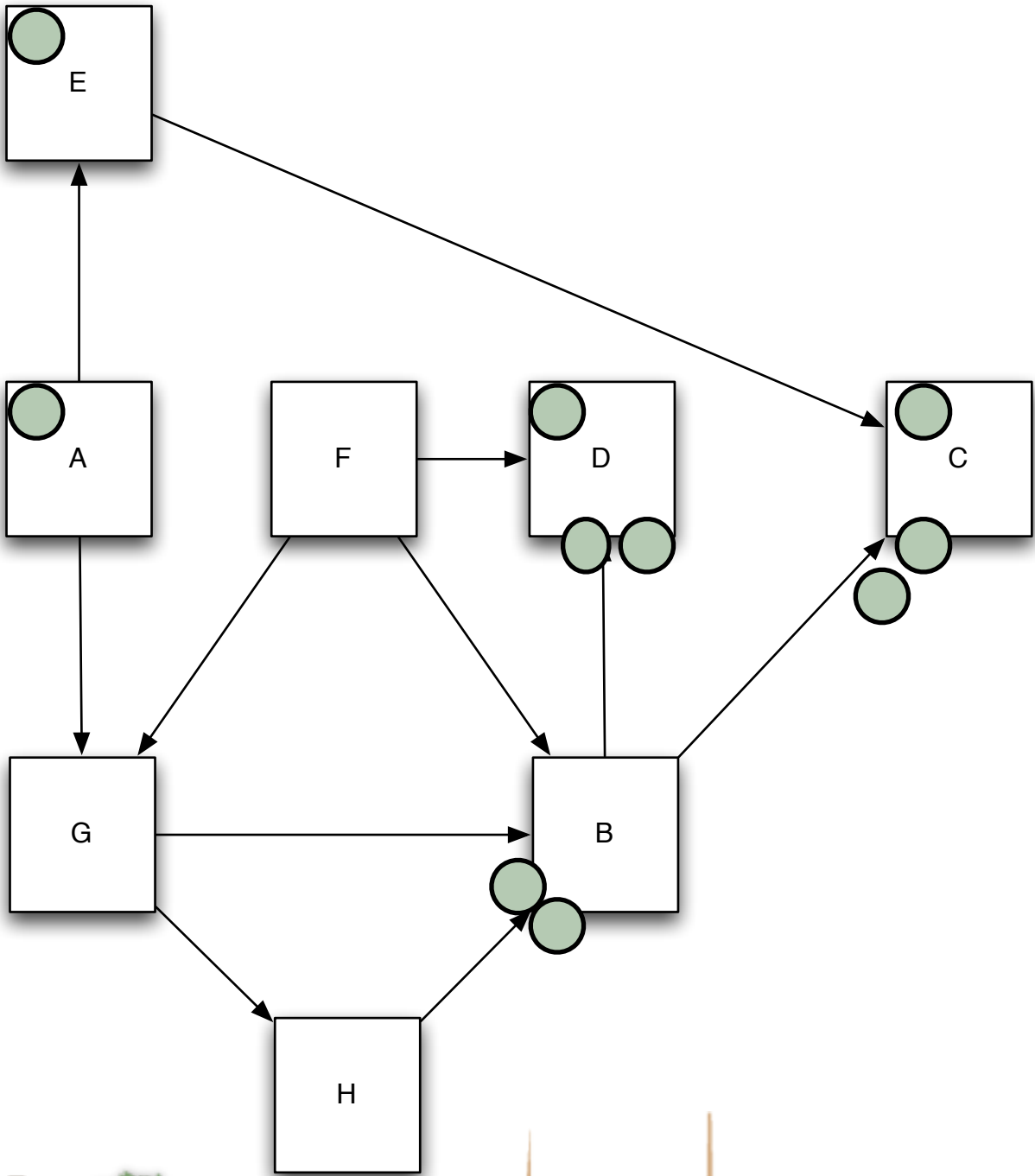
Long-Term visit rate

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0



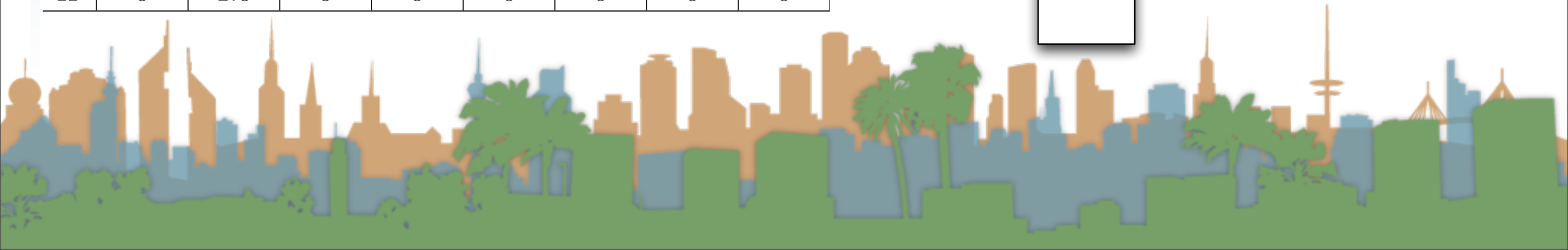
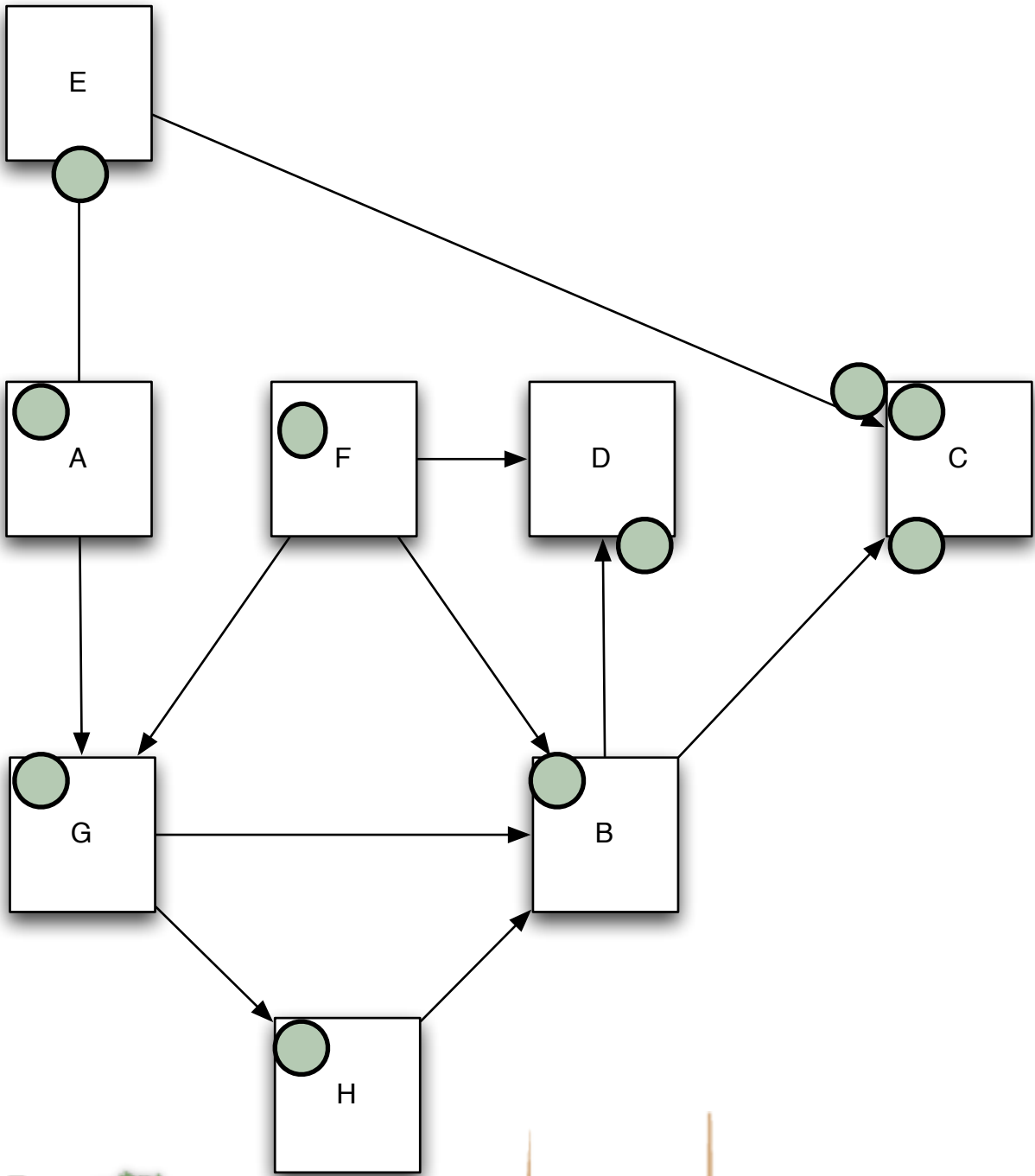
Long-Term visit rate

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0



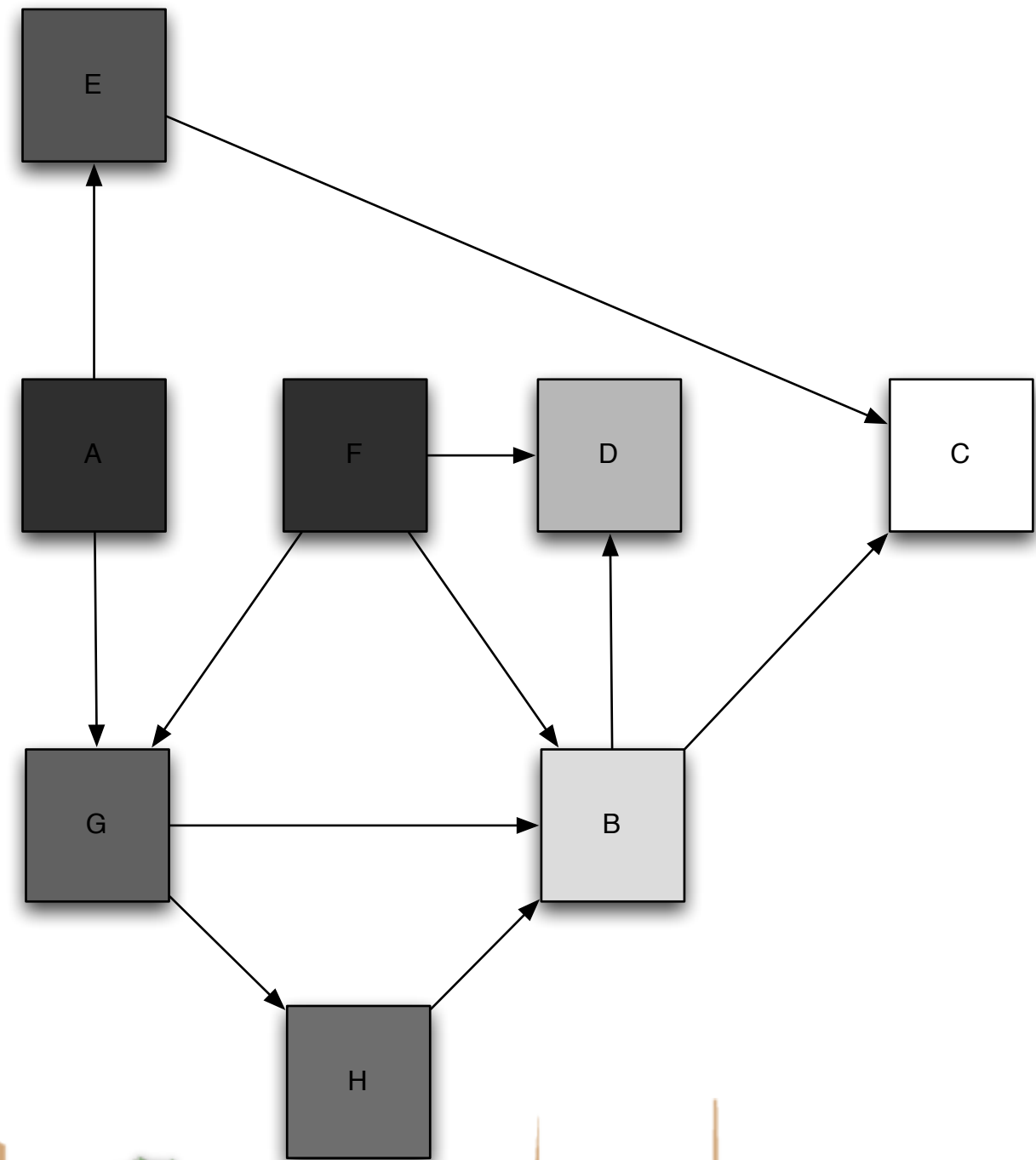
Long-Term visit rate

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0



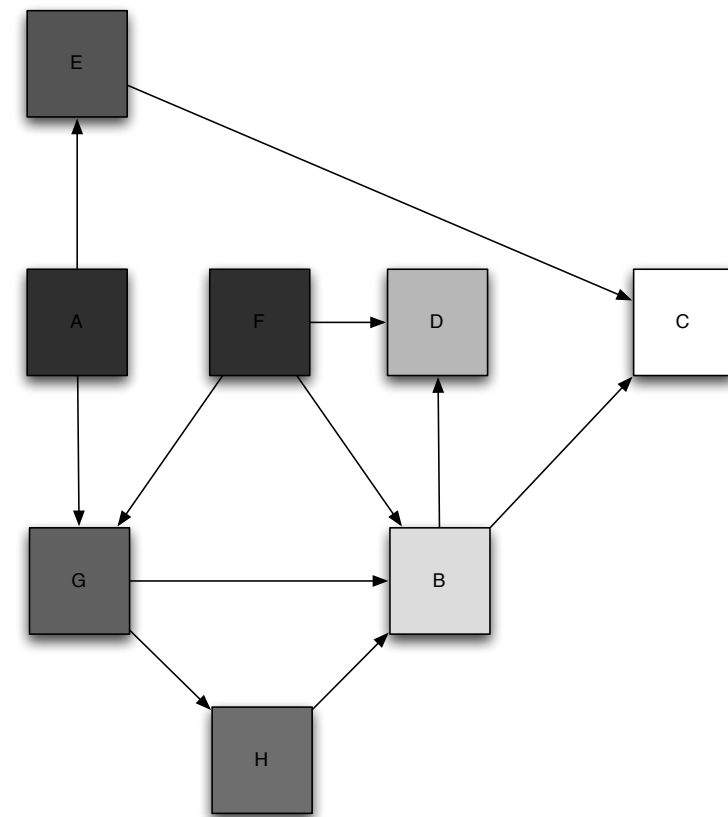
## Long-Term visit rate

- A: 5%
- B: 21%
- C: 23%
- D: 18%
- E: 8%
- F: 5%
- G: 9%
- H: 10%



## Some properties of Markov chains

- **Ergodic:**
  - All states can reach all states
  - What did we have to do to enable this for a web graph?
- **Steady State Theorem:**
  - Every ergodic markov chain has a steady state -> has a PageRank

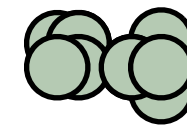


## Calculating PageRank

- Visual representation to math representation

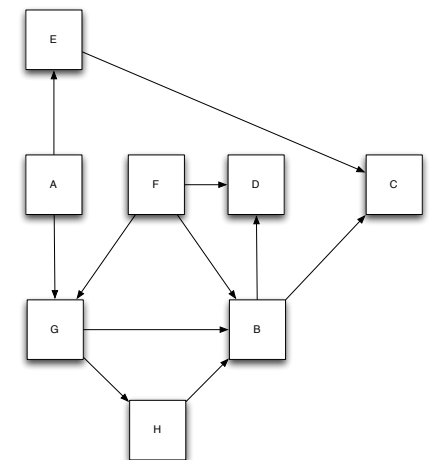
$\vec{x}_0$

<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
1.0	0	0	0	0	0	0	0



$P$

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0





## Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_0 P = \vec{x}_1$$

<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
1.0	0	0	0	0	0	0	0

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0



## Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_0 P = \vec{x}_1$$

1.0	A
0	B
0	C
0	D
0	E
0	F
0	G
0	H

	A	B	C	D	E	F	G	H
A	0	0	0	0	0.5	0	0.5	0
B	0	0	0.5	0.5	0	0	0	0
C	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
D	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
E	0	0	1.0	0	0	0	0	0
F	0	0.33	0	0.33	0	0	0.33	0
G	0	0.5	0	0	0	0	0	0.5
H	0	1.0	0	0	0	0	0	0

0
---



## Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_0 P = \vec{x}_1$$

	A	B	C	D	E	F	G	H
A	0	0	0	0	0.5	0	0.5	0
B	0	0	0.5	0.5	0	0	0	0
C	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
D	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
E	0	0	1.0	0	0	0	0	0
F	0	0.33	0	0.33	0	0	0.33	0
G	0	0.5	0	0	0	0	0	0.5
H	0	0	0	0	0	0	0	0

0	0
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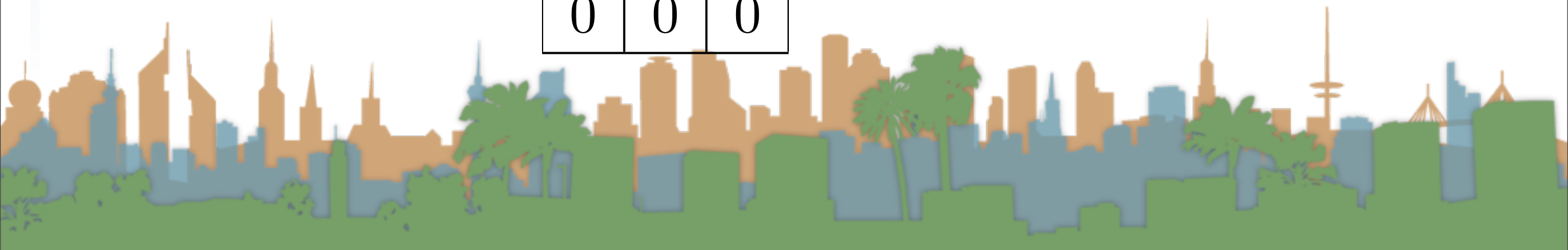
## Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_0 P = \vec{x}_1$$

	A	B	C	D	E	F	G	H
A	0	0	0	0	0.5	0	0.5	0
B	0	0	0.5	0.5	0	0	0	0
C	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
D	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
E	0	0	0	0	0	0	0	0
F	0	0.33	0	0.33	0	0	0.33	0
G	0	0.5	0	0	0	0	0	0.5
H	0	1.0	0	0	0	0	0	0

0	0	0
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## Calculating PageRank

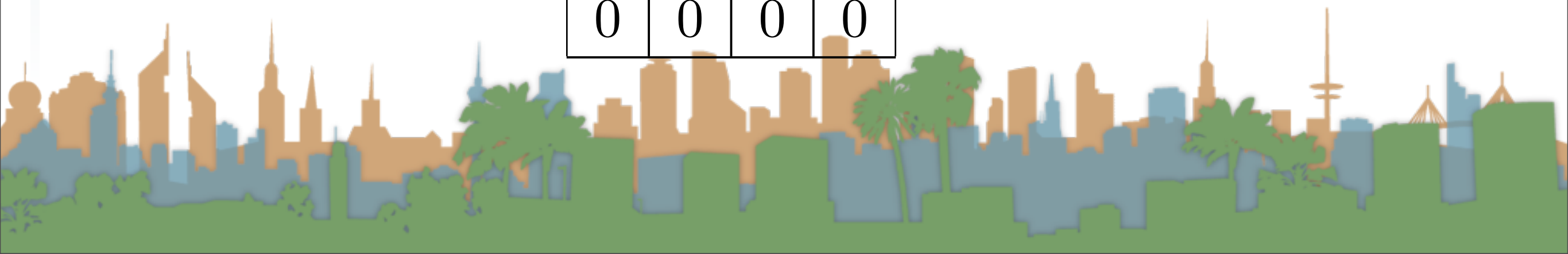
- Take one step is multiplying state vector

$$\vec{x}_0 P = \vec{x}_1$$

times transition probability matrix

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0

0	0	0	0
---	---	---	---



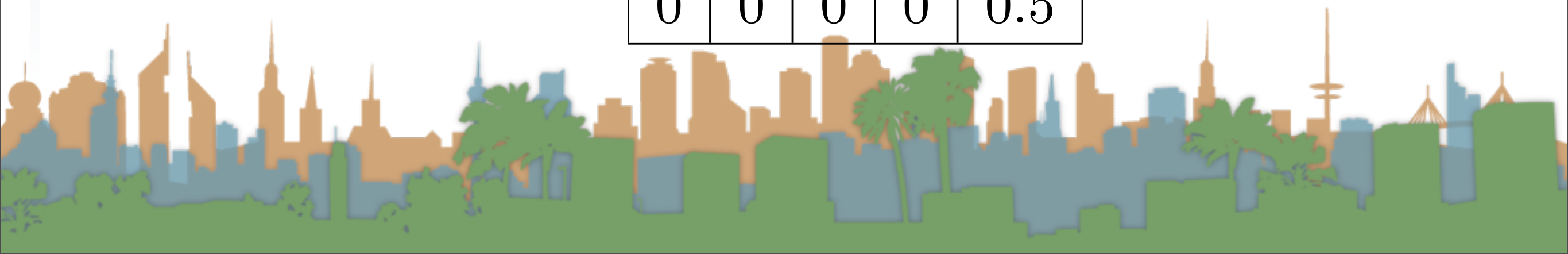
## Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_0 P = \vec{x}_1$$

	A	B	C	D	E	F	G	H
A	0	0	0	0	0.5	0	0.5	0
B	0	0	0.5	0.5	0	0	0	0
C	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
D	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
E	0	0	1.0	0	0	0	0	0
F	0	0.33	0	0.33	0	0	0.33	0
G	0	0.5	0	0	0	0	0	0.5
H	0	1.0	0	0	0	0	0	0

0	0	0	0	0.5
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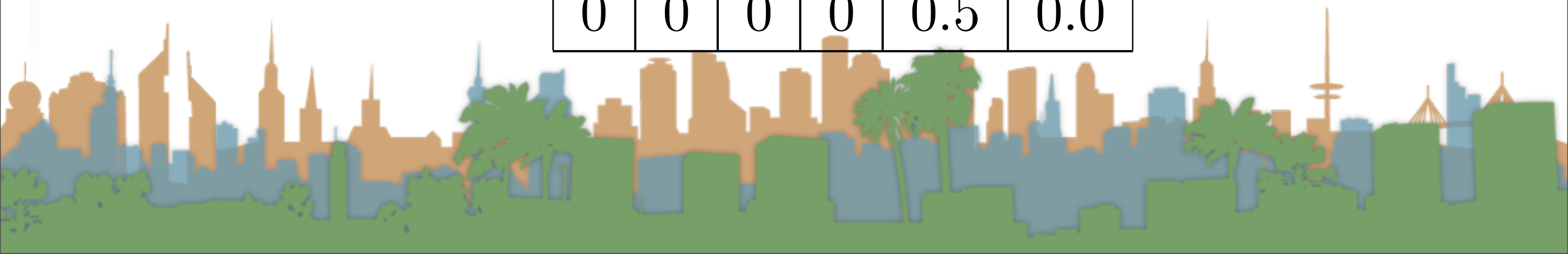
## Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_0 P = \vec{x}_1$$

	A	B	C	D	E	F	G	H
A	0	0	0	0	0.5	0	0.5	0
B	0	0	0.5	0.5	0	0	0	0
C	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
D	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
E	0	0	1.0	0	0	0	0	0
F	0	0.33	0	0.33	0	0	0.33	0
G	0	0.5	0	0	0	0	0	0.5
H	0	1.0	0	0	0	0	0	0

0	0	0	0	0.5	0.0
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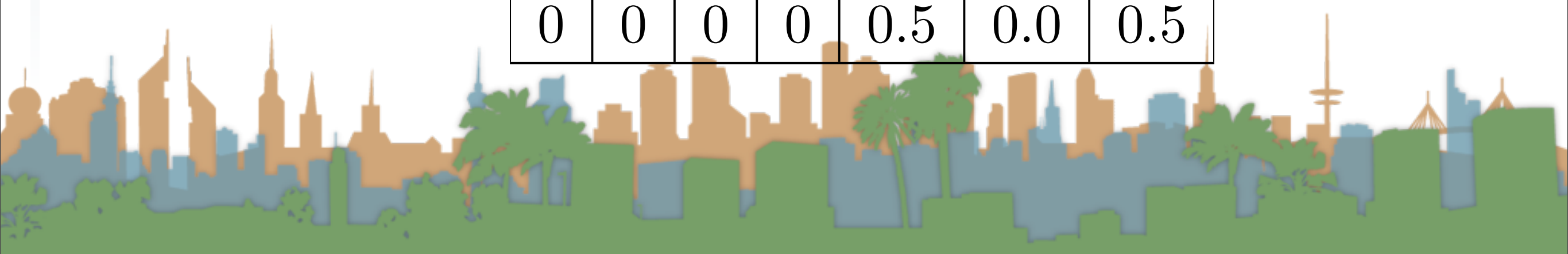
## Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_0 P = \vec{x}_1$$

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0

0	0	0	0	0.5	0.0	0.5
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## Calculating PageRank

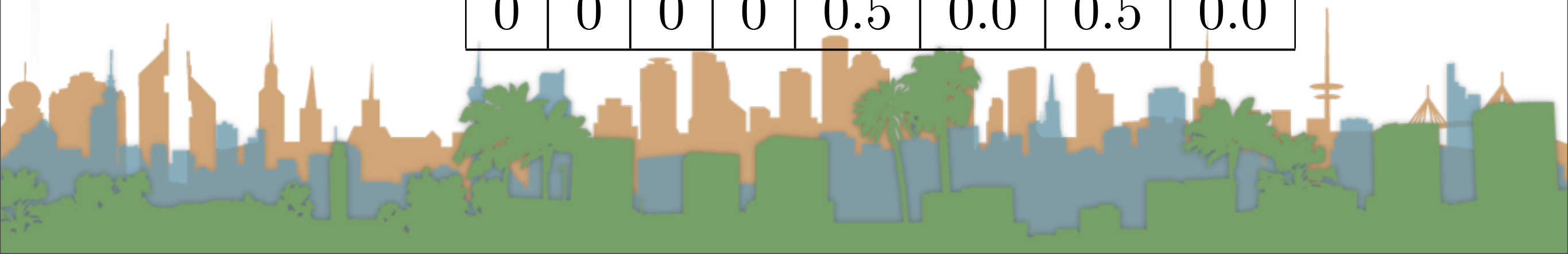
- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_0 P = \vec{x}_1$$

1.0	A
0	B
0	C
0	D
0.125	E
0.125	F
0	G
0.5	H

	A	B	C	D	E	F	G
A	0	0	0	0	0.5	0	0.5
B	0	0	0.5	0.5	0	0	0
C	0.125	0.125	0.125	0.125	0.125	0.125	0.125
D	0.125	0.125	0.125	0.125	0.125	0.125	0.125
E	0	0	1.0	0	0	0	0
F	0	0.33	0	0.33	0	0	0.33
G	0	0.5	0	0	0	0	0
H	0	1.0	0	0	0	0	0

0	0	0	0	0.5	0.0	0.5	0.0
---	---	---	---	-----	-----	-----	-----



## Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_0 P = \vec{x}_1$$

<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
1.0	0	0	0	0	0	0	0

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0

$$\vec{x}_1 = \begin{array}{|c|c|c|c|c|c|c|c|c|} \hline 0 & 0 & 0 & 0 & 0.5 & 0 & 0.5 & 0 \\ \hline \end{array}$$



## Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_1 P = \vec{x}_2$$

0	0	0	0	0.5	0	0.5	0
---	---	---	---	-----	---	-----	---

	<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>
<i>A</i>	0	0	0	0	0.5	0	0.5	0
<i>B</i>	0	0	0.5	0.5	0	0	0	0
<i>C</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>D</i>	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
<i>E</i>	0	0	1.0	0	0	0	0	0
<i>F</i>	0	0.33	0	0.33	0	0	0.33	0
<i>G</i>	0	0.5	0	0	0	0	0	0.5
<i>H</i>	0	1.0	0	0	0	0	0	0

$$\vec{x}_2 =$$

0	0.25	0.5	0	0.0	0	0.0	0.25
---	------	-----	---	-----	---	-----	------



## Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_1 P = \vec{x}_2$$

0	0.25	0.5	0	0.0	0	0.0	0.25		$A$	$B$	$C$	$D$	$E$	$F$	$G$	$H$
								$A$	0	0	0	0	0.5	0	0.5	0
								$B$	0	0	0.5	0.5	0	0	0	0
								$C$	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
								$D$	0.125	0.125	0.125	0.125	0.125	0.125	0.125	0.125
								$E$	0	0	1.0	0	0	0	0	0
								$F$	0	0.33	0	0.33	0	0	0.33	0
								$G$	0	0.5	0	0	0	0	0	0.5
								$H$	0	1.0	0	0	0	0	0	0

$$\vec{x}_3 = \begin{bmatrix} 0.0625 & 0.3125 & 0.1875 & 0.1875 & 0.0625 & 0.0625 & 0.0625 & 0.0625 \end{bmatrix}$$



# Calculating PageRank

- Take one step is multiplying state vector times transition probability matrix

$$\vec{x}_1 P = \vec{x}_2$$

$$\lim_{(n \rightarrow \infty)} x_n = PageRank$$



## Long-Term visit rate

- A: 5%
- B: 21%
- C: 23%
- D: 18%
- E: 8%
- F: 5%
- G: 9%
- H: 10%

