

Assignment 6 - Markov Chains*

Due: Monday, April 30

This is optional, and will be averaged into your grade **only** if it improves your grade

Overview

In this assignment you will explore different approaches to analyzing Markov chains.

You will use two data sets for this assignment:

- <http://www.cs.utah.edu/~jeffp/teaching/cs5955/A6/M.dat>
- <http://www.cs.utah.edu/~jeffp/teaching/cs5955/A6/L.dat>

These data sets are in matrix format and can be loaded into MATLAB or OCTAVE. By calling

`load filename` (for instance `load M.dat`)

it will put in memory the the data in the file, for instance in the above example the matrix M . You can then display this matrix by typing

`M`

As usual, it is highly recommended that you use LaTeX for this assignment. If you do not, you may lose points if your assignment is difficult to read or hard to follow. Find a sample form in this directory: <http://www.cs.utah.edu/~jeffp/teaching/latex/>

1 Finding q_* (12 points)

We will consider four ways to find $q_* = M^t q_0$ as $t \rightarrow \infty$.

Matrix Power: Choose some large enough value t , and create M^t . Then apply $q_* = (M^t)q_0$. There are two ways to create M^t , first we can just let $M^{i+1} = M^i * M$, repeating this process $t - 1$ times. Alternatively, (for simplicity assume t is a power of 2), then in $\log_2 t$ steps create $M^{2^i} = M^{2^{i-1}} * M^{2^{i-1}}$.

State Propagation: Iterate $q_{i+1} = M * q_i$ for some large enough number t iterations.

Random Walk: Starting with a fixed state $q_0 = [00 \dots 1 \dots 00]^T$ where there is only a 1 at the i th entry, and then transition to a new state with only a 1 in the i' th entry by choosing a new location proportional to the values in the i th column of M . Iterate this some large number t_0 of steps to get state q'_0 . (This is the *burn in period*.)

Now make t new step starting at q'_0 and record the location after each step. Keep track of how many times you have recorded each location and estimate q_* as the normalized version (recall $\|q_*\|_1 = 1$) of the vector of these counts.

Eigen-Analysis: Compute `eig(M)` and take the first eigenvector after it has been normalized.

A (4 points): Run each method (with $t = 500$, $q_0 = [100 \dots 0]^T$ and $t_0 = 20$ when needed) and report the answers.

B (2 points): Rerun the Matrix Power and State Propagation techniques with $q_0 = [0.1, 0.1, \dots, 0.1]^T$. For what value of t is required to get as close to the true answer as the older initial state?

C (2 points): For what value of t does the Random Walk approach need to get an answer as accurate as the State Propagation with $t = 500$? This is a randomized algorithm so you will want to complete multiple trials in order to be confident in your answer.

D (4 points): Explain at least one **Pro** and one **Con** of each approach. The **Pro** should explain a situation when it is the best option to use. The **Con** should explain why another approach may be better for some situation.

2 Taxation (4 points)

Repeat the trials in part **A** above using taxation $\beta = 0.85$ so at each step, with probability $1 - \beta$, any state jumps to a random node. It is useful to see how the outcome changes with respect to the results from Question 1. Recall that this output is the *PageRank* vector of the graph represented by M .

Briefly explain (no more than 2 sentences) what you needed to do in order to alter the process in question 1 to apply this taxation.

3 BONUS (6 points)

A (3 points): Consider the transition probability matrix L . Starting at state 10, how many steps in a random walk do you need (in expectation) to reach state 1 for the first time? This quantity is known as the *hitting time*.

B (3 points): L is known as a *lolly pop graph*. It can be generalized to a graph with n nodes where there are $n/2$ nodes that form a clique, and then $n/2$ nodes which forms a path that connects to the clique at one end.

Given a starting state in the clique, explain asymptotically in n (i.e. using big-Oh notation) how many steps you expect you need to take to reach the node at the end of the path not connected to the clique.