

Project 3 Markov Chains

Jacob Hopkins

4/16/2020

Contents

2.1 Markov Chains	1
-----------------------------	---

2.1 Markov Chains

A Markov chain is a discrete-time discrete-state process.

Given three states with values thus:

State 0: 0.64 0.32 0.04

State 1: 0.40 0.50 0.10

State 2: 0.25 0.50 0.25

We start at state 0. Then we sample the states possibilities based upon the the probabilities provided in the matrix, and this bring us to a new state.

Recording how many times we go to a given state and evaluating how many states we went to total can give us the probability of each state in total.

We then compare them to theoretical results.

$\pi = (a, b, c)$

$$0.64a + 0.32b + 0.04c = a$$

$$0.40a + 0.50b + 0.10c = b$$

$$0.25a + 0.50b + 0.25c = c$$

or equivalent to:

$$0.32b + 0.04c = 0.36a$$

$$0.40a + 0.10c = 0.50b$$

$$0.25a + 0.50b = 0.75c$$

multiply the first by 5 and the second by 2

$$1.6b + 0.20c = 1.8a$$

$$0.80a + 0.20c = b$$

$$0.25a + 0.50b = 0.75c$$

subtracting the two equations first and second we get:

$$1.6b - 0.8a = 1.8a - b$$

simplifying we get: $2.6a = 2.6b$

or

$$a = b$$

using this in the third equation we get:

$$0.25a + 0.50a = 0.75c$$

$$0.75a = 0.75c$$

$$a = c$$

so

$$a = b = c = 1$$

so

$$\text{pi} = (0.33, 0.33, 0.33)$$

Unless I made a mistake in calculating this

```

state_0_probability = 0.3333
state_1_probability = 0.3333
state_2_probability = 0.3333

p_00 = 0.64
p_01 = 0.32
p_02 = 0.04
p_10 = 0.40
p_11 = 0.50
p_12 = 0.10
p_20 = 0.25
p_21 = 0.50
p_22 = 0.25

P = matrix(c(p_00, p_01, p_02, p_10, p_11, p_12, p_20, p_21, p_22), nrow=3, byrow=TRUE)

number_of_transitions = 10000

current_state = 0

states_population = c(0,1,2)

states_count = c(0,0,0)

print(P)

##           State 0 State 1 State 2
## State 0      0.64    0.32    0.04
## State 1      0.40    0.50    0.10
## State 2      0.25    0.50    0.25

for (current_transition in 1:number_of_transitions){

  for (states in states_population){
    if(current_state == states){

```

```

        states_count[states] = states_count[states] + 1
        current_state = sample(states_population, 1, replace = FALSE, prob = P[states+1])
        break
    }
}

}

cat("\n")

cat("Probability of State 0: ", states_count[0]/number_of_transitions * 100.0, "%\n")

## Probability of State 0:   %

cat("Probability of State 1: ", states_count[1]/number_of_transitions * 100.0, "%\n")

## Probability of State 1:  40.96 %

cat("Probability of State 2: ", states_count[2]/number_of_transitions * 100.0, "%\n")

## Probability of State 2:   8.49 %

cat("\n")

cat("Theoretical Probability of State 0: ", state_0_probability, "%\n")

## Theoretical Probability of State 0:  0.3333 %

cat("Theoretical Probability of State 1: ", state_1_probability, "%\n")

## Theoretical Probability of State 1:  0.3333 %

cat("Theoretical Probability of State 2: ", state_2_probability, "%\n")

## Theoretical Probability of State 2:  0.3333 %

cat("\n")

cat("Distance of Simulated to Theoretical Probability of State 0: ", 100 * (states_count[0] - state_0_probability), "%\n")

## Distance of Simulated to Theoretical Probability of State 0:   %

```

```
cat("Distance of Simulated to Theoretical Probability of State 1: ", 100 * (states_con  
## Distance of Simulated to Theoretical Probability of State 1: 7.63 %  
cat("Distance of Simulated to Theoretical Probability of State 2: ", 100 * (states_con  
## Distance of Simulated to Theoretical Probability of State 2: -24.84 %
```