

L21: Markov Chains

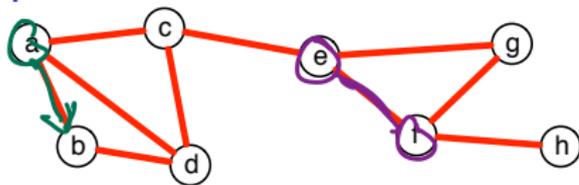
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April 7, 2025

Markov Chain : Life Lessons

- ▶ **[L1]** *Only your current position matters going forward, don't worry about the past.*
- ▶ **[L2]** *You just need to worry about one step at a time; you will get there eventually (or you won't).*
- ▶ **[L3]** *In the limit, everyone has perfect karma.*

Graphs



Mathematically: $G = (V, E)$ where

$V = \{a, b, c, d, e, f, g\}$ and

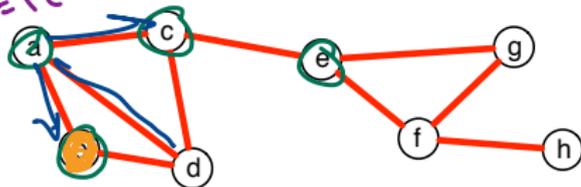
$E = \left\{ \{a, b\}, \{a, c\}, \{a, d\}, \{b, d\}, \{c, d\}, \{c, e\}, \{e, f\}, \{e, g\}, \{f, g\}, \{f, h\} \right\}$.

Matrix-Style: As a matrix with 1 if there is an edge, and 0 otherwise.
(For a directed graph, it may not be symmetric).

$$G = \begin{matrix} & \begin{matrix} a & b & c & d & e & f & g & h \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \\ f \\ g \\ h \end{matrix} & \begin{bmatrix} 0 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \end{matrix} = \begin{matrix} & \begin{matrix} a & b & c & d & e & f & g & h \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \\ f \\ g \\ h \end{matrix} & \begin{bmatrix} 0 & 1 & 1 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 1 & 0 & 0 & 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 1 & 1 \\ 0 & 0 & 0 & 0 & 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 \end{bmatrix} \end{matrix}$$

Markov Chain

$$P_i = P(a \rightarrow \cdot)$$



$\{V, P, q\}$: V node set, P probability transition matrix, q initial state q_0

e.g. $q^T = [0 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]$ or $q^T = [0.1 \ 0 \ 0 \ 0 \ 0.3 \ 0 \ 0.6 \ 0 \ 0]$.

$$P = \begin{pmatrix} 0 & 1/2 & 1/3 & 1/3 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 1/3 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 1/3 & 1/3 & 0 & 0 & 0 \\ 1/3 & 1/2 & 1/3 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/3 & 0 & 0 & 1/3 & 1/2 & 0 \\ 0 & 0 & 0 & 0 & 1/3 & 0 & 1/2 & 1 \\ 0 & 0 & 0 & 0 & 1/3 & 1/3 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1/3 & 0 & 0 \end{pmatrix}$$

$P_i = \Delta_n$
probability
vector

$$P_{ij} = \text{Prob}(\dots j) A_i$$

$$P_i = \frac{1}{|A_i|} A_i$$

$$\|A_i\|_1 = \left(\sum_{j=1}^n |A_{ij}| \right)^{1/2}$$

Transitions

$$P = \begin{pmatrix} 0 & 1/2 & 1/3 & 1/3 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 1/3 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 1/3 & 1/3 & 0 & 0 & 0 \\ 1/3 & 1/2 & 1/3 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/3 & 0 & 0 & 1/3 & 1/2 & 0 \\ 0 & 0 & 0 & 0 & 1/3 & 0 & 1/2 & 1 \\ 0 & 0 & 0 & 0 & 1/3 & 1/3 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1/3 & 0 & 0 \end{pmatrix} \text{ and } q^T = [0 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]$$

$$q_1 = [Pq] = \begin{bmatrix} \frac{1}{2} & 0 & 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \end{bmatrix}^T$$

Transitions

$$P = \begin{pmatrix} 0 & 1/2 & 1/3 & 0 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 1/3 & 1/3 & 0 & 0 & 0 \\ 1/3 & 1/2 & 1/3 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/3 & 0 & 0 & 1/3 & 1/2 & 0 \\ 0 & 0 & 0 & 0 & 1/3 & 0 & 1/2 & 1 \\ 0 & 0 & 0 & 0 & 1/3 & 1/3 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1/3 & 0 & 0 & 0 \end{pmatrix} \text{ and } q^T = [0 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]$$

one step P

$$q_1 = Pq = \begin{bmatrix} \frac{1}{2} & 0 & 0 & \frac{1}{2} & 0 & 0 & 0 & 0 \end{bmatrix}^T.$$

$$q_2 = Pq_1 = PPq = P^2q = \begin{bmatrix} \frac{1}{6} & \frac{2}{6} & \frac{2}{6} & \frac{1}{6} & 0 & 0 & 0 & 0 \end{bmatrix}^T.$$

Transitions

elems normalized.

$$P = \begin{pmatrix} 0 & 1/2 & 1/3 & 1/3 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 1/3 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 1/3 & 1/3 & 0 & 0 & 0 \\ 1/3 & 1/2 & 1/3 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/3 & 0 & 0 & 1/3 & 1/2 & 0 \\ 0 & 0 & 0 & 0 & 1/3 & 0 & 1/2 & 1 \\ 0 & 0 & 0 & 0 & 1/3 & 1/3 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1/3 & 0 & 0 \end{pmatrix} \text{ and } q^T = [0 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]$$

$$q_1 = Pq = \left[\frac{1}{2} \ 0 \ 0 \ \frac{1}{2} \ 0 \ 0 \ 0 \ 0 \right]^T.$$

$$q_2 = Pq_1 = PPq = P^2q = \left[\frac{1}{6} \ \frac{2}{6} \ \frac{2}{6} \ \frac{1}{6} \ 0 \ 0 \ 0 \ 0 \right]^T.$$

$$q_3 = Pq_2 = \left[\frac{1}{3} \ \frac{1}{9} \ \frac{1}{9} \ \frac{1}{3} \ \frac{1}{9} \ 0 \ 0 \ 0 \right]^T.$$

In the limit: $q_n = P^n q = P(P \dots (P q) \dots)$

NOT

uniform
uniform?

$$q_n = \left(\frac{1}{n}, \frac{1}{n}, \dots, \frac{1}{n} \right)$$

Transitions

$$P = \begin{pmatrix} 0 & 1/2 & 1/3 & 1/3 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 1/3 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 0 & 1/3 & 1/3 & 0 & 0 & 0 \\ 1/3 & 1/2 & 1/3 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/3 & 0 & 0 & 1/3 & 1/2 & 0 \\ 0 & 0 & 0 & 0 & 1/3 & 0 & 1/2 & 1 \\ 0 & 0 & 0 & 0 & 1/3 & 1/3 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1/3 & 0 & 0 \end{pmatrix} \text{ and } q^T = [0 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]$$

$$q_1 = Pq = \left[\frac{1}{2} \ 0 \ 0 \ \frac{1}{2} \ 0 \ 0 \ 0 \ 0 \right]^T.$$

$$q_2 = Pq_1 = PPq = P^2q = \left[\frac{1}{6} \ \frac{2}{6} \ \frac{2}{6} \ \frac{1}{6} \ 0 \ 0 \ 0 \ 0 \right]^T.$$

$$q_3 = Pq_2 = \left[\frac{1}{3} \ \frac{1}{9} \ \frac{1}{9} \ \frac{1}{3} \ \frac{1}{9} \ 0 \ 0 \ 0 \right]^T.$$

In the limit: $q_n = P^n q$

[L1] *Only your current position matters going forward,
don't worry about the past.*

Two ways to think of Markov Chain

(1) Only consider 1 possible state (node) at a time
(e.g. $g_i = [0, 0, 1, 0, 0, 0]$)

(2) Probability Distribution on states (nodes)
(e.g. $g_0 = [\frac{1}{3}, 0, 0, \frac{1}{6}, 0, \frac{1}{2}]$)

Limiting state g_n as n goes to ∞
in limit
 $g_n = P^n g_0$

Ergodic (Markov Chain)

(When limit exists)

MC is ergodic if $\exists t$ so that
for all $n \geq t$ $g_n(j) > 0$.
for all $j \in V$

Not ergodic iff

- ① has absorbing & transient states
- ② disconnected.
- ③ cyclic.

Absorbing and Transient Examples

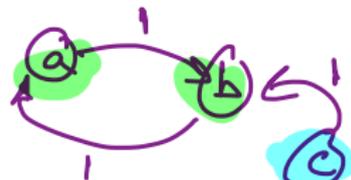
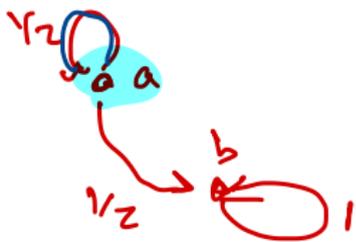
$$P_a^n \left(\frac{1}{2}^n \right)$$



$$\begin{matrix} & \begin{matrix} a & b \end{matrix} \\ \begin{matrix} a \\ b \end{matrix} & \begin{pmatrix} 1/2 & 0 \\ 1/2 & 1 \end{pmatrix} \end{matrix}$$

$$\begin{matrix} & \begin{matrix} a & b & c \end{matrix} \\ \begin{matrix} a \\ b \\ c \end{matrix} & \begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 1 \\ 0 & 0 & 0 \end{pmatrix} \end{matrix}$$

$$\begin{pmatrix} & \begin{matrix} a & b & c & d & e & f \end{matrix} \\ \begin{matrix} a \\ b \\ c \\ d \\ e \\ f \end{matrix} & \begin{pmatrix} 1/2 & 1/2 & 0 & 0 & 0 & 0 \\ 1/2 & 49/100 & 0 & 0 & 0 & 0 \\ 0 & 1/100 & 1/4 & 1/4 & 1/4 & 1/4 \\ 0 & 0 & 1/4 & 1/4 & 1/4 & 1/4 \\ 0 & 0 & 1/4 & 1/4 & 1/4 & 1/4 \\ 0 & 0 & 1/4 & 1/4 & 1/4 & 1/4 \end{pmatrix} \end{pmatrix}$$



No absorbing + transient
it can go from any
node to any other.

Unconnected Examples

$$\begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\begin{pmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

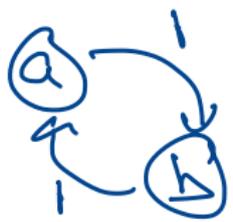
$$\begin{pmatrix} 1/2 & 1/2 & 0 & 0 & 0 & 0 \\ 1/2 & 1/2 & 0 & 0 & 0 & 0 \\ 0 & 0 & 1/3 & 1/2 & 1/3 & 0 \\ 0 & 0 & 1/3 & 0 & 1/3 & 0 \\ 0 & 0 & 1/3 & 1/2 & 1/3 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}$$



Cyclic Examples



most common \rightarrow



$$\begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

$$g_0 = [1, 0]$$

$$\begin{pmatrix} 0 & 1 & 0 \\ 0 & 0 & 1 \\ 1 & 0 & 0 \end{pmatrix}$$

$$g_1 = P g_0 = [0, 1]$$

$$g_{17} = [0, 1]$$



$$\begin{pmatrix} 0 & 1/2 & 1/2 & 1/2 & 1/2 & 0 \\ 1/4 & 0 & 0 & 0 & 0 & 1/4 \\ 1/4 & 0 & 0 & 0 & 0 & 1/4 \\ 1/4 & 0 & 0 & 0 & 0 & 1/4 \\ 1/4 & 0 & 0 & 0 & 0 & 1/4 \\ 0 & 1/2 & 1/2 & 1/2 & 1/2 & 0 \end{pmatrix}$$

0.005
 \uparrow
 breaks

cyclic structure.

Limiting State

if MC ergodic

Let $P^* = P^n$ as $n \rightarrow \infty$.

Let $q_* = P^*q$.

initial state q_0 does not matter.

Limiting State

Let $P^* = P^n$ as $n \rightarrow \infty$.

Let $q_* = P^*q$.

[L2] *You just need to worry about one step at a time;
you will get there eventually (or you won't).*

if has
(might not) absorbing and transient states

Delicate Balance

limiting state

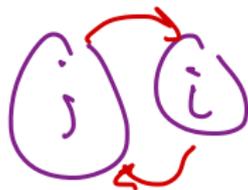
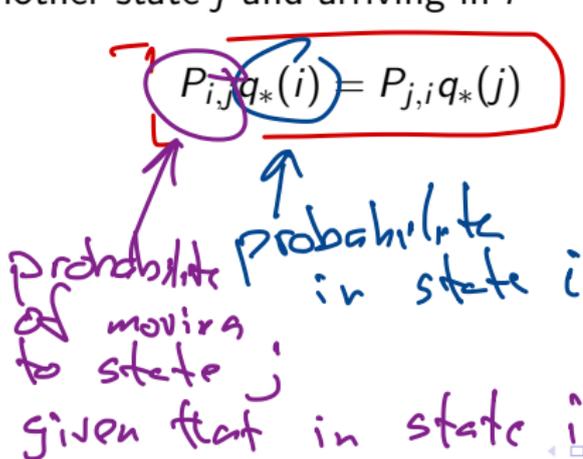
$$q_{\infty} \in \Delta_{|V|}$$

Let $P^* = P^n$ as $n \rightarrow \infty$.

Let $q_* = P^* q$.

Also $q_* = \underbrace{P}_{q_*} \underbrace{P^* q}_{q_*}$ thus $q_* = P q_*$.

So the probability of being in a state i and leaving to j is the same as being in another state j and arriving in i



Delicate Balance

$$q_* = P q_*$$

q_* eigenvector of P .

Let $P^* = P^n$ as $n \rightarrow \infty$.

Let $q_* = P^* q$.

Also $q_* = P P^* q$ thus $q_* = P q_*$.

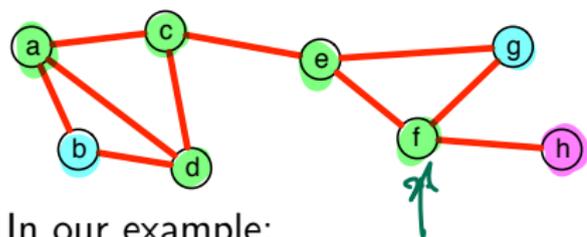
So the probability of being in a state i and leaving to j is the same as being in another state j and arriving in i

$$P_{i,j} q_*(i) = P_{j,i} q_*(j)$$

[L3] *In the limit, everyone has perfect karma.*

Limiting State

not
uniform



In our example:

$$q_* = (0.15, 0.1, 0.15, 0.15, 0.15, 0.15, 0.1, 0.05)$$

$$= \left(\frac{3}{20}, \frac{1}{10}, \frac{3}{20}, \frac{3}{20}, \frac{3}{20}, \frac{3}{20}, \frac{1}{10}, \frac{1}{20} \right)$$

f

Algorithms for computing g_x

① $\text{eig}(P) \Rightarrow$ top eigenvectors $v_i \in \mathbb{S}^{n-1}$
 $g_x = \frac{v_i}{\|v_i\|}$

② $g_n = P(P \dots P(g_0) \dots)$ n steps
 n large P
Power method.

③ $g_n = P^n g_0$ for large enough n
 $n = 2^k$ $P^2 \rightarrow P^4 \rightarrow P^8 \rightarrow \dots \rightarrow P^{2^k}$

④ Random walk on P
keep track of where I walk.

Metropolis Algorithm

Metropolis, Rosenbluth, Rosenbluth, Teller, and Teller in 1953

Metropolis Algorithm

Metropolis, Rosenbluth, Rosenbluth, Teller, and Teller in 1953

Metropolis on V and w

Initialize $v_0 = [0 \ 0 \ 0 \ \dots \ 1 \ \dots \ 0 \ 0]^T$.

repeat

 Generate $u \sim K(v, \cdot)$

if ($w(u) \geq w(v_i)$) **then**

 Set $v_{i+1} = u$

else

 With probability $w(u)/w(v)$ set $v_{i+1} = u$

else

 Set $v_{i+1} = v_i$

until “converged”

return $V = \{v_1, v_2, \dots, \}$

