

Markov Chains

- Consider a set of states $S = \{1, 2, \dots, k\}$
- Assume states evolve over time in discrete time steps
- Let $X_n =$ state at time $n = 0, 1, 2, \dots$

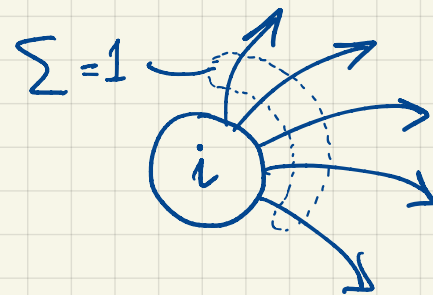
We say we have a Markov Chain iff $(m < n)$:

$$P(X_n = j \mid x_0, x_1, x_2, \dots, x_m = i) = P(X_n = j \mid x_m = i)$$

(Conditioning on x_m , makes x_n independent of x_0, \dots, x_{m-1})

Furthermore, when $m = n-1$, above probability is given by P_{ij} and is called the transition prob. from state i to state j

$$\sum_j P_{ij} = 1$$



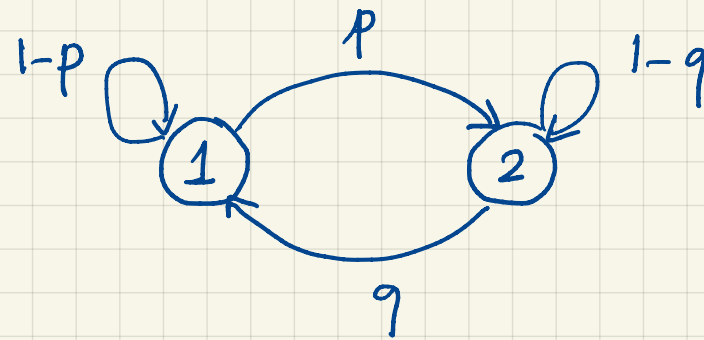
Matrix Representation

Here's a simple Markov chain given as a matrix and illustrated by a state diagram

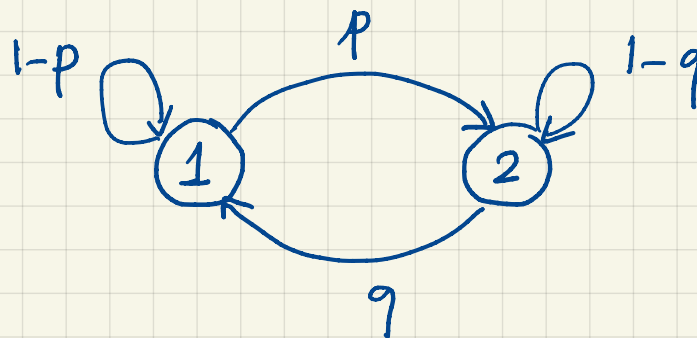
$$P = \begin{bmatrix} 1-p & p \\ q & 1-q \end{bmatrix}$$

All the transition probabilities

$$P_{11} = 1-p \quad P_{12} = p \quad P_{21} = q \quad P_{22} = 1-q$$



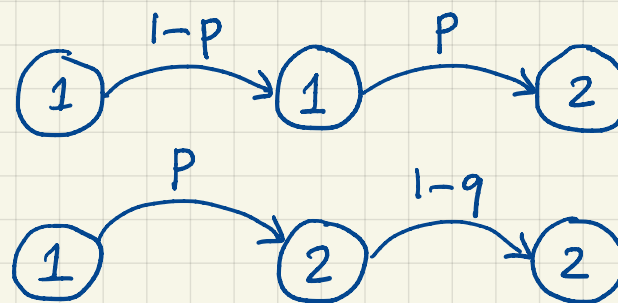
$$P = \begin{bmatrix} 1-p & p \\ q & 1-q \end{bmatrix}$$



Let's do some calculations: What's $P(X_2=2 \mid X_0=1)$?

"Probability of being in state 2 at time 2
given we are in state 1 at time 0"

Visual interpretation:



$$(1-p)p + p(1-q)$$

We have "visually" used the Markov property. (see next)

$$P(X_2=2 | X_0=1) = \sum_i P(X_2=2 | X_1=i, X_0=1) P(X_1=i | X_0=1)$$

$$= P(X_2=2 | X_1=1, X_0=1) P(X_1=1 | X_0=1) + P(X_2=2 | X_1=2, X_0=1) P(X_1=2 | X_0=1)$$

$$= P(X_2=2 | X_1=1) P(X_1=1 | X_0=1) + P(X_2=2 | X_1=2) P(X_1=2 | X_0=1)$$

$$= p(1-p) + (1-q)p$$

Observation: This is the entry in first row second column of P^2

$$P^2 = \begin{bmatrix} 1-p & p \\ q & 1-q \end{bmatrix} \times \begin{bmatrix} 1-p & p \\ q & 1-q \end{bmatrix} = \begin{bmatrix} - & (1-p)p + p(1-q) \\ - & - \end{bmatrix}$$

In general $P(X_n=j | X_m=i) = P_{ij}^{n-m}$ for $n > m$.

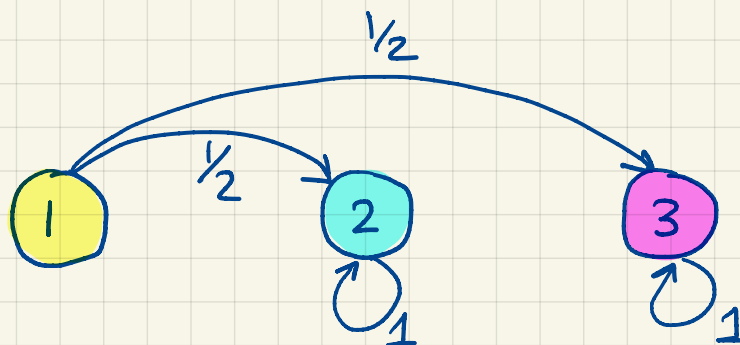
Two Properties

Irreducible and aperiodic

- The matrix P is irreducible iff:

$$\forall i, j. \exists n. P(X_n = j \mid X_0 = i) > 0$$

- In terms of the state diagram, this means there is a path from i to j (not necessarily an immediate transition)
- The previous Markov Chain is irreducible.
- The one below is NOT.



$$\begin{bmatrix} 0 & 1/2 & 1/2 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

What's important about irreducibility?

Irreducible $\Rightarrow \exists$ unique stationary distribution of states

Unique Vector $\pi \neq 0$ ($\sum \pi_i = 1$)

$$\pi P = \pi$$

Why called stationary?

$$[\pi_1 \ \pi_2 \ \dots \ \pi_i \ \dots \ \pi_k] \begin{bmatrix} p_{1i} \\ p_{2i} \\ \vdots \\ p_{ki} \end{bmatrix} = [\dots \ \pi_i \ \dots]$$

$$\sum_j \pi_j p_{ji} = \pi_i$$

$$\pi_1 P_{1i} + \pi_2 P_{2i} + \dots + \pi_i P_{ii} + \dots + \pi_k P_{ki} = \pi_i$$

- If π is the prob. distribution over states at time n , then the left hand side is the prob. of being in state i at time $n+1$.
- If this is π_i , then the prob. of state i has not changed.
- This is true for all i .

$$P = \begin{bmatrix} 1-p & p \\ q & 1-q \end{bmatrix}$$

This is irreducible if $p, q > 0$

$$\pi = \left[\frac{q}{p+q}, \frac{p}{p+q} \right]$$

$$P = \begin{bmatrix} 0 & \frac{1}{2} & \frac{1}{2} \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

This is not irreducible

any $\pi = [0 \ p \ 1-p]$ is stationary

(not unique)

Is it possible to have no stationary distribution? Not if

we have a finite set of states. There has to be at least one.

How do we find the stationary dist.?

Solve a system of linear equations.

Example: $[\pi_1, \pi_2] \begin{bmatrix} 1-p & p \\ q & 1-q \end{bmatrix} = [\pi_1, \pi_2]$

① $\pi_1(1-p) + \pi_2 q = \pi_1$

$\pi_1 p + \pi_2(1-q) = \pi_2$ (same equation!)

add: $\pi_1 + \pi_2 = 1 \Rightarrow \pi_2 = 1 - \pi_1$ ②

① $\Rightarrow \pi_1(1-p) + (1-\pi_1)q = \pi_1$

$\pi_1(1-p) - \pi_1 q - \pi_1 = -q$

$\pi_1(\cancel{1-p} - q - \cancel{1}) = -q \Rightarrow \pi_1(p+q) = q \Rightarrow \pi_1 = \frac{q}{p+q}$

k equations, 1 for each $i \in \{1, 2, \dots, k\}$

$$i: \sum_{j=1}^k \pi_j P_{ji} = \pi_i$$

If we sum the first $k-1$ equations, we obtain the last

$$\sum_{i=1}^{k-1} \sum_{j=1}^k \pi_j P_{ji} = \sum_{i=1}^{k-1} \pi_i$$

$$\sum_{j=1}^k \sum_{i=1}^{k-1} \pi_j P_{ji} = \sum_{i=1}^{k-1} \pi_i$$

$$\sum_{j=1}^k \pi_j \sum_{i=1}^{k-1} P_{ji} = \sum_{i=1}^{k-1} \pi_i$$

$$\sum_{j=1}^k \pi_j (1 - P_{jk}) = \sum_{i=1}^{k-1} \pi_i$$

$$\sum_{j=1}^k \pi_j - \sum_{j=1}^k \pi_j P_{jk} = \sum_{i=1}^{k-1} \pi_i$$

$$\pi_k - \sum_{j=1}^k \pi_j P_{jk} = 0$$

$$\sum_{j=1}^k \pi_j P_{jk} = \pi_k \quad (k^{\text{th}} \text{ equation})$$

In general, solve:

$$\begin{bmatrix} \pi_1 & \pi_2 & \dots & \pi_{k-1} & \pi_k \end{bmatrix} \begin{bmatrix} \vdots \\ \vdots \\ \vdots \\ \vdots \\ \vdots \end{bmatrix} = \begin{bmatrix} \pi_1 & \pi_2 & \dots & \pi_{k-1} & 1 \end{bmatrix}$$

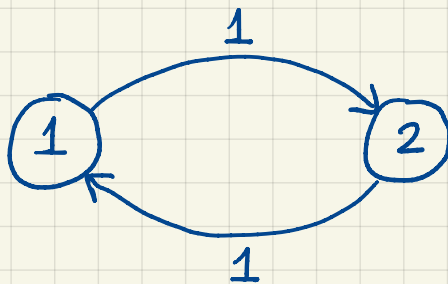
P

The matrix P is aperiodic iff:

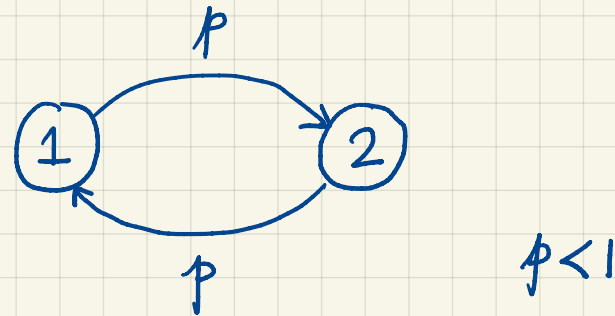
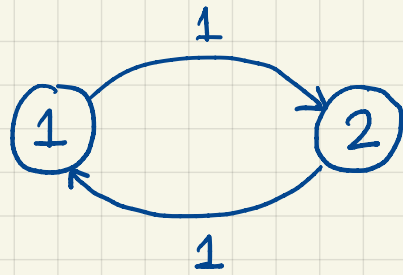
$$\forall i. \gcd \{n : P(X_n = i \mid X_0 = i) > 0\} = 1$$

What does that mean? If we consider the lengths of all cycles that brings us back to state i , those lengths have no common divisor (except 1).

This chain is irreducible, but NOT aperiodic



Let's compare:



Both are irreducible with $\pi = [\frac{1}{2} \ \frac{1}{2}]$ as the stationary dist.

What if we start in the states with prob. $[\frac{1}{3} \ \frac{2}{3}]$?

What will happen next ?

$$\begin{bmatrix} \frac{1}{3} & \frac{2}{3} \end{bmatrix} \underbrace{\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}}_{\text{periodic}} = \begin{bmatrix} \frac{2}{3} & \frac{1}{3} \end{bmatrix}$$

$$\begin{bmatrix} \frac{2}{3} & \frac{1}{3} \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} \frac{1}{3} & \frac{2}{3} \end{bmatrix}$$

No convergence to $[\frac{1}{2} \ \frac{1}{2}]$

$$\begin{bmatrix} q & 1-q \end{bmatrix} \underbrace{\begin{bmatrix} 1-p & p \\ p & 1-p \end{bmatrix}}_{\text{aperiodic}} = \begin{bmatrix} \underbrace{q(1-p) + (1-q)p}_{q'} & \underbrace{qp + (1-q)(1-p)}_{1-q'} \end{bmatrix}$$

$$\begin{bmatrix} q' & 1-q' \end{bmatrix} \begin{bmatrix} 1-p & p \\ p & 1-p \end{bmatrix} = \begin{bmatrix} q'' & 1-q'' \end{bmatrix} \dots$$

At the limit, we are going to converge to $\begin{bmatrix} 1/2 & 1/2 \end{bmatrix}$

Lesson: • If you start the system in the stationary dist., you will stay in the stationary dist.

• If you start the system arbitrarily, you may not reach the stationary dist. But you will if aperiodic!

For most practical purposes, we only need irreducibility.

- If a Markov chain is irreducible, then

$$\frac{1}{n+1} \sum_{t=0}^n f(x_t) \xrightarrow{n \rightarrow \infty} \sum_i \pi_i f(x_i)$$

- In addition, if chain is aperiodic, then

$$\lim_{n \rightarrow \infty} P^n = \begin{bmatrix} \pi_1 & \dots & \pi_k \\ \pi_1 & \dots & \pi_k \\ \vdots & & \\ \pi_1 & \dots & \pi_k \end{bmatrix}$$

$$\lim_{n \rightarrow \infty} P(X_n = i | X_0) = \pi_i$$

↑
No effect