

# Markov Chain I

July 28, 2022

- HW 8 is optional

- HW 7 is the last graded hw

# Example 1a

Toss an unfair coin  $\mathbb{P}(\text{Head}) = p$  for  $N$  times. What's the fraction of time for observing heads out of all outcomes?

$$\mathbb{E}(\#H) = N \cdot p$$

$$\#H \sim \text{Binomial}(N, p)$$

$$\text{fraction of } H = \frac{Np}{N} = p$$

# Example 1b

Now we have two unfair coins, each is biased to either head or tail.

Coin 1:  $\mathbb{P}(\text{Head}) = p$ ; Coin 2:  $\mathbb{P}(\text{Head}) = 1 - p$ .

If seeing head, then use coin 1 for next toss; if seeing tail, then use coin 2 for the next toss.

Toss  $N$  times, what's the fraction of time for observing heads out of all outcomes?

# Markov Chain (Discrete Time Finite MC):

## State Space:

At each time step  $n$ , the state is denoted by  $X_n$ . The collection of all possible values a state can take is called the state space  $S = \{1, 2, \dots, K\}$  for a finite number  $K$ .

## Transition Probability:

$$0 \leq P_{ij} = \mathbb{P}(X_{n+1} = j \mid X_n = i) \quad \forall i, j \in S$$

*next step*      *current*  
↑                    ↑

$$\sum_{j=1}^K P_{ij} = 1, \quad \forall i$$

$P_{ij}$  or  $P(i, j)$  forms a matrix  $P \in \mathbb{R}^{K \times K}$   
 $P(i, j) = P_{ij}$

# Markov Property

$$\mathbb{P}(\underline{X_{n+1} = j} \mid \underline{X_n = i, X_{n-1} = i_{n-1}, \dots, X_0 = i_0})$$

$$= \mathbb{P}(\underline{X_{n+1} = j} \mid \underline{X_n = i}) = P_{ij}$$

# Example 1b

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If seeing head, then use coin 1 for next toss; if seeing tail, then use coin 2 for the next toss.

Toss N times, what's the fraction of time for observing heads out of all outcomes?

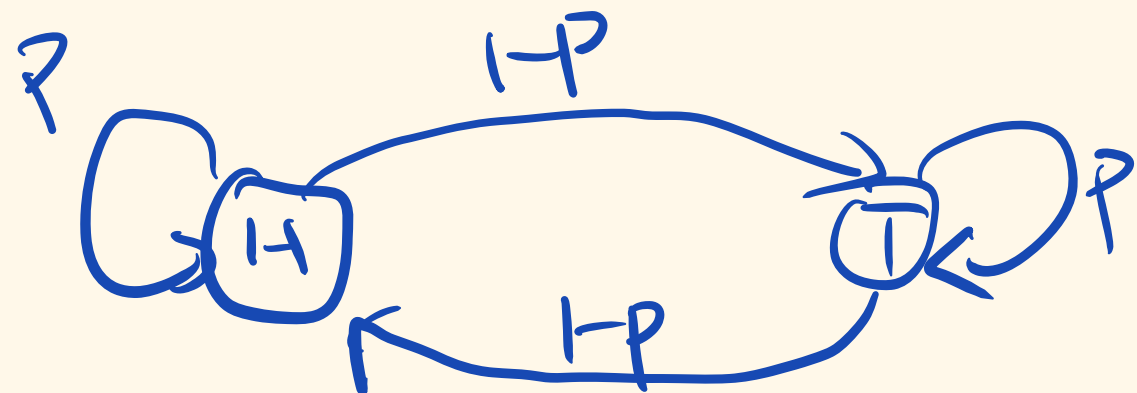
$$S = \{1, 2\}$$

$$\text{or } S = \{H, T\}$$

1: Head. 2: Tail

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

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



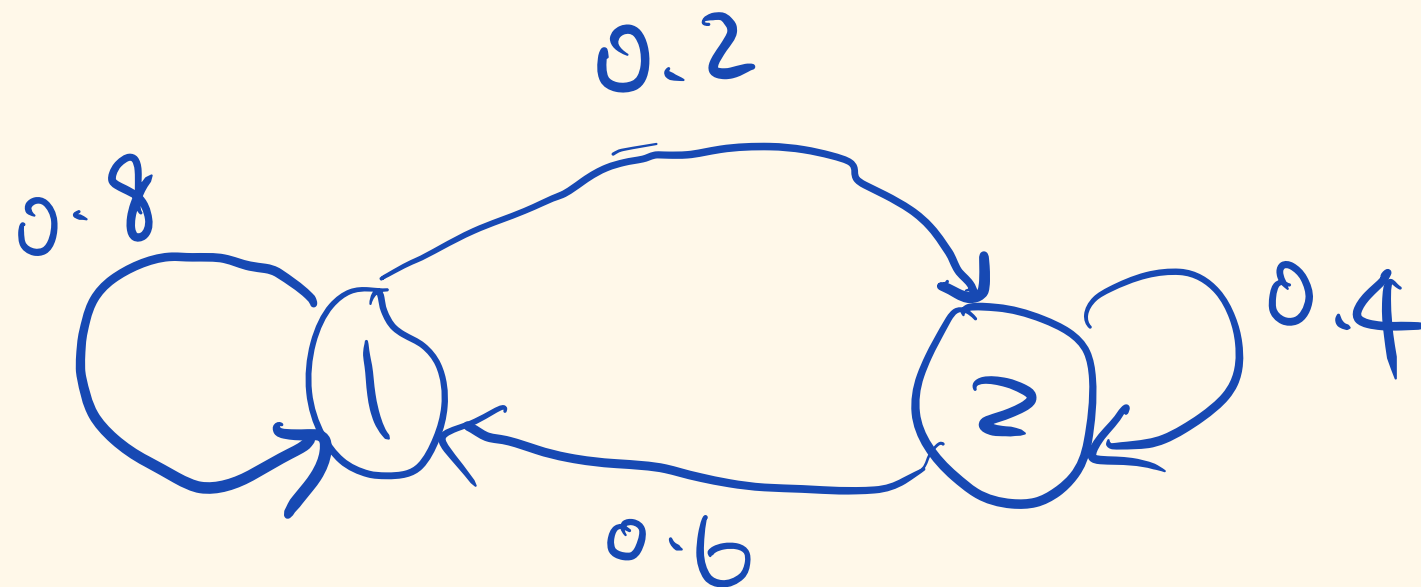
$$P = \begin{bmatrix} p & 1-p \\ 1-p & p \end{bmatrix}$$

# Example 2: Alice takes probability class.

Alice is either (1) up-to-date or (2) fall behind

$$P_{11} = 0.8, P_{12} = 0.2, P_{21} = 0.6, P_{22} = 0.4$$

$$P = \begin{bmatrix} 0.8 & 0.2 \\ 0.6 & 0.4 \end{bmatrix}$$





Probability of being in state  $j$ , at time step  $n$

$$\pi_n(j) := P(\Sigma_n = j) \quad \text{e.g. } \pi_n = [\pi_n(1), \pi_n(2)]$$

$$\pi_n(j) = P(\Sigma_n = j) = \sum_{i=1}^k P(\Sigma_n = j | \Sigma_{n-1} = i) P(\Sigma_{n-1} = i)$$

$$= \sum_{i=1}^k P_{ij} \pi_{n-1}(i) \quad \forall j \in S$$

$$\begin{array}{c} \pi_n = \pi_{n-1} P \\ \text{---} = \text{---} \begin{array}{|c|} \hline \\ \hline \end{array} \\ 1 \times k \quad 1 \times k \quad k \times k \end{array}$$

$\pi_0$  : initial distribution

$$\pi_0 = [\pi_0(1), \pi_0(2), \dots, \pi_0(k)]$$

for  $S = \{1, \dots, k\}$ .

$$\pi_n = \pi_{n-1} P = \pi_{n-2} P^2 = \dots = \pi_0 P^n$$

# Balance Equation

A distribution  $\pi$  is invariant for the transition probability  $P$  if it satisfies the balance equation

$$\pi P = \pi$$

$$\sum_j \pi(j) = 1$$

does not depend on a specific time point.

has many names

$\pi$ : invariant dist, stationary dist, steady-state prob.

Thm: if  $\pi_n = \pi_0, \forall n \geq 0 \Leftrightarrow \pi_0$  is invariant.

$\pi$  is a row vector.

$$\pi_n(i) := P(\sum_{k=1}^n X_k = i)$$

Example 2: Alice takes probability class.

Alice is either (1) up-to-date or (2) fall behind. Find the stationary distribution.

$$P_{11} = 0.8, P_{12} = 0.2, P_{21} = 0.6, P_{22} = 0.4$$

$$P = \begin{bmatrix} 0.8 & 0.2 \\ 0.6 & 0.4 \end{bmatrix}$$

$$\pi = \pi P \Rightarrow \begin{cases} \pi_1 = \pi_1 P_{11} + \pi_2 P_{21} \\ \pi_2 = \pi_1 P_{12} + \pi_2 P_{22} \end{cases}$$

$$\begin{cases} \pi_1 = 0.8\pi_1 + 0.6\pi_2 \\ \pi_2 = 0.2\pi_1 + 0.4\pi_2 \end{cases} \Rightarrow \begin{cases} \pi_1 = 3\pi_2 \\ \pi_1 + \pi_2 = 1 \end{cases} \Rightarrow \begin{cases} \pi_1 = 0.75 \\ \pi_2 = 0.25 \end{cases}$$

$$\ast \pi = \pi P, \pi = \pi P^2, \pi = \pi P^3, \pi(i) \geq 0, \sum_i \pi(i) = 1$$

# Properties of Markov Chain

Q1: Does an invariant distribution always exist?

Q2: Is it unique?

Q3: Does  $\pi_n$  approach an invariant distribution?

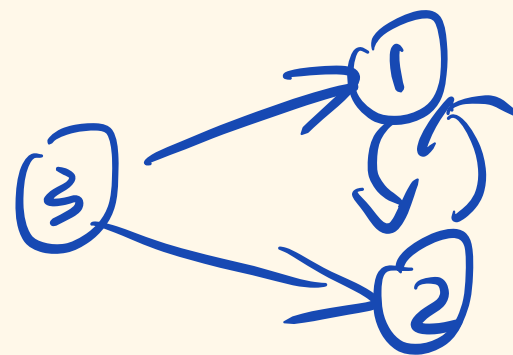
$$\pi_n(j) = \mathbb{P}(X_n = j)$$

# Properties of Markov Chain

## Irreducibility

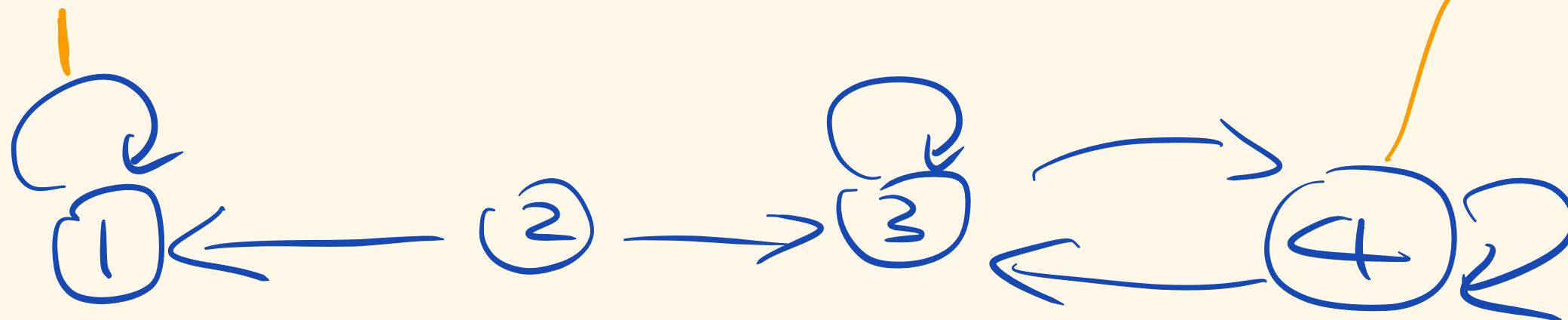
A Markov chain is irreducible if it can go from any state to any other state, possibly after many steps.

Ex.



reducible

Ex. :



↓  
reducible.

# Properties of Markov Chain

## (a) periodicity

for an irreducible Markov Chain defined on state space  $S$  with transition probability  $P$ .

Let

$$d(i) := \gcd\{n \geq 1 \mid P^n(i, i) > 0\}$$

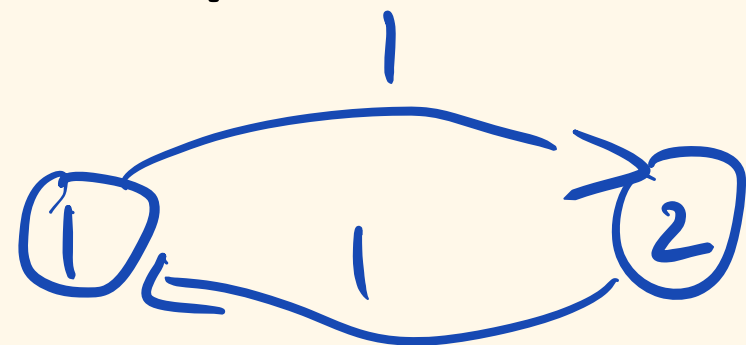
Then,  $d(i)$  has some value for all  $i$ ,  $d(i) = \underline{d}$ . *← see proof in the end*

If  $d=1$ , MC is aperiodic

If  $d > 1$ , MC is periodic with period  $d$ .

*\* the period of states is the same as the period of this MC. if MC is irreducible.*

# Example 3a



• irreducible.

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

• periodic.  $d=2$

$$P = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \quad P^2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \Rightarrow P^3 = P$$

$$\pi = (0.5, 0.5) \leftarrow \text{invariant distribution}$$

$$\pi_n = (1, 0) \quad \pi_{n+1} = (0, 1) \quad , \quad \pi_{n+2} = (1, 0)$$

$$\lim_{n \rightarrow \infty} \pi_n \quad \times$$

$$\pi_n \rightarrow \pi \quad n \rightarrow \infty$$

# Example 3b: Alice studies Markov Chain

Find the ~~invariant distribution~~ of Alice's study status.

$$P = \begin{bmatrix} 0.8 & 0.2 \\ 0.6 & 0.4 \end{bmatrix} \quad P^n(i, i) > 0 \quad \forall n$$

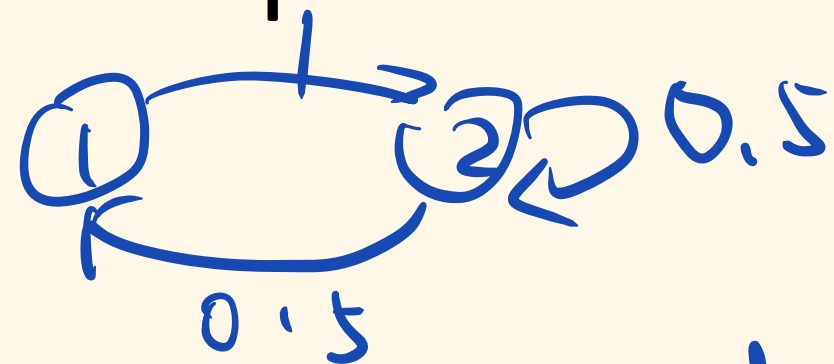
$d=1$ , aperiodic

$$P, P^2, P^3, \dots \quad \underline{P^n = \begin{bmatrix} 0.75 & 0.25 \\ 0.75 & 0.25 \end{bmatrix}}$$

$$\pi P = \pi \quad \Rightarrow \quad \pi = [0.75, 0.25]$$



Example 3c: Find the invariant distribution.



$$P = \begin{bmatrix} 0 & 1 \\ 0.5 & 0.5 \end{bmatrix}, \quad P^2 = \begin{bmatrix} 0.5 & 0.5 \\ 0.25 & 0.75 \end{bmatrix} \dots$$

$$d=1$$

$$\pi = \pi P \Rightarrow \pi = [0.5, 0.5]$$

A(n) irreducible Markov Chain with self-loop is aperiodic.

note: the reverse is not true;  
counter example: random walk on a triangle.

# Theorem for Markov Chain

1) If a Markov Chain is finite and irreducible:

it has unique invariant distribution  $\pi$  and.

does not matter if  
the MC is periodic  
or not

$\pi(i)$  long term fraction of time.

$$\lim_{N \rightarrow \infty} \frac{1}{N} \sum_{n=0}^{N-1} \mathbb{I} \{ \{ \bar{X}_n = i \} \} = \pi(i)$$

← invariant distribution

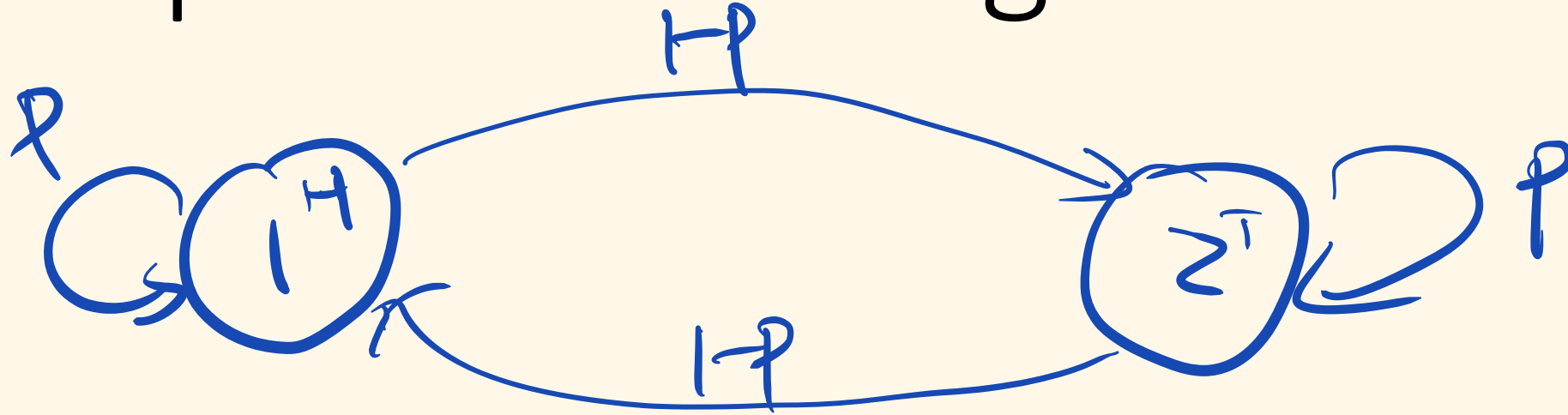
2) If this Markov Chain is also aperiodic:

$$\text{then } \pi_n \rightarrow \pi, n \rightarrow \infty$$

$$\pi_n, \pi_{n+1}, \pi_{n+2} = \pi_n, \pi_{n+3} = \pi_{n+1}$$

$d=2$

Example 1b. Tossing two unfair coins

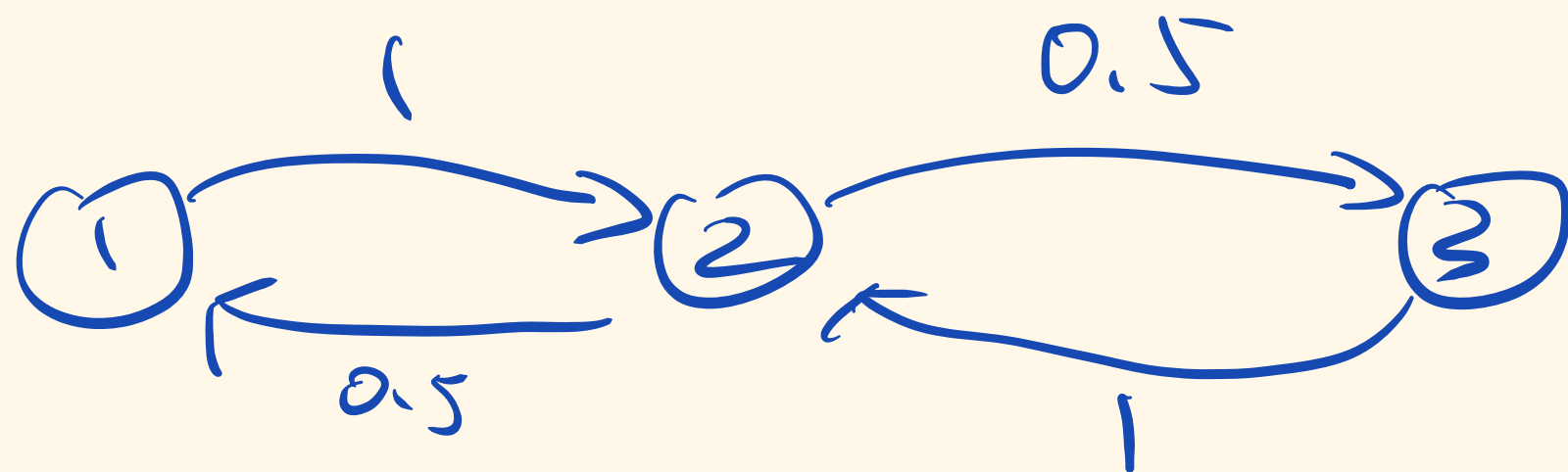


$$\pi = [0.5, 0.5]$$

One more example

$$P = \begin{bmatrix} 0 & 1 & 0 \\ 0.5 & 0 & 0.5 \\ 0 & 1 & 0 \end{bmatrix}$$

$$d=2$$



$P^2, P^3 = P$  this is a periodic MC, but we can still define the stationary distribution  $\pi$

$$\pi = \pi P \Rightarrow \pi = \left[ \frac{1}{4}, \frac{1}{2}, \frac{1}{4} \right], \text{ long term fraction of time at state 2 is } \frac{1}{2}.$$

For people who are interested, this is not required for exams.

lemma : for an irreducible MC.

$d(i) := \gcd \{ n \geq 1 \mid P^n(i, i) > 0 \}$  has same value

for all state  $i$ .

proof : pick  $\forall j \neq i$ , we show that  $d(j) \leq d(i)$

and the flip  $j \rightarrow i$ ,  $d(i) \leq d(j)$  so  $d(i) = d(j)$

MC is irreducible, so  $\exists m, p^m(j, i) > 0$  &  $\exists n, p^n(i, j) > 0$

now by definition of  $d(i)$

$\exists N$ , such that  $p^N(i, i) > 0$  and  $p^{N+d(i)}(i, i) > 0$

$p^{m+N+n}(j, j) > 0$  and  $p^{m+N+d(i)+n}(j, j) > 0$

$\Rightarrow$   $m+N+n$ ,  $m+N+d(i)+n$   $\in \{n \geq 1 \mid p^n(j, j) > 0\}$

so  $d(j) \leq d(i)$  because  $\gcd\{k, k+d(i), \dots\} \leq d(i)$

reverse  $i, j$ , do it again we have  $d(i) \leq d(j)$

so  $d(i) = d(j) \quad \forall i \neq j$



