

# Markov Chain II

Aug 1, 2022

# Markov Chain review

- Markov Property
- State Space
- Transition Probability
- Balance Equation
  - Invariant distribution (aka, stationary distribution, steady state distribution)
- Irreducible MC has unique Invariant distribution
- Irreducible and aperiodic MC has  $\pi_n \rightarrow \pi, n \rightarrow \infty$
- Long- term fraction of time of state  $i$

$$\lim_{N \rightarrow \infty} \left\{ \frac{1}{N} \mathbb{I}\{\mathbf{X}_n = i\} \right\} = \pi(i)$$

Given an irreducible MC, if it contains self loop, then it is aperiodic

- The reserve is not true.
  - counterexample, random walk on a triangle

```
P =  
  
      0      0.5000      0.5000  
0.5000      0      0.5000  
0.5000      0.5000      0
```

```
>> P^2  
  
ans =  
  
      0.5000      0.2500      0.2500  
      0.2500      0.5000      0.2500  
      0.2500      0.2500      0.5000
```

```
>> P^4  
  
ans =  
  
      0.3750      0.3125      0.3125  
      0.3125      0.3750      0.3125  
      0.3125      0.3125      0.3750
```

```
>> P^6  
  
ans =  
  
      0.3438      0.3281      0.3281  
      0.3281      0.3438      0.3281  
      0.3281      0.3281      0.3438
```

```
>> P^3  
  
ans =  
  
      0.2500      0.3750      0.3750  
      0.3750      0.2500      0.3750  
      0.3750      0.3750      0.2500
```

```
>> P^5  
  
ans =  
  
      0.3125      0.3438      0.3438  
      0.3438      0.3125      0.3438  
      0.3438      0.3438      0.3125
```

```
>> P^7  
  
ans =  
  
      0.3281      0.3359      0.3359  
      0.3359      0.3281      0.3359  
      0.3359      0.3359      0.3281
```

```
>> P^100  
  
ans =  
  
      0.3333      0.3333      0.3333  
      0.3333      0.3333      0.3333  
      0.3333      0.3333      0.3333
```

Given an irreducible MC, if it is aperiodic, then  $\pi_n \rightarrow \pi, n \rightarrow \infty$

- The reserve is not true
  - Counterexample: random walk on a square

```
P =  
  
    0    0.5000    0    0.5000  
0.5000    0    0.5000    0  
    0    0.5000    0    0.5000  
0.5000    0    0.5000    0
```

```
>> P^2  
  
ans =  
  
    0.5000    0    0.5000    0  
    0    0.5000    0    0.5000  
    0.5000    0    0.5000    0  
    0    0.5000    0    0.5000
```

```
>> P^3  
  
ans =  
  
    0    0.5000    0    0.5000  
0.5000    0    0.5000    0  
    0    0.5000    0    0.5000  
0.5000    0    0.5000    0
```

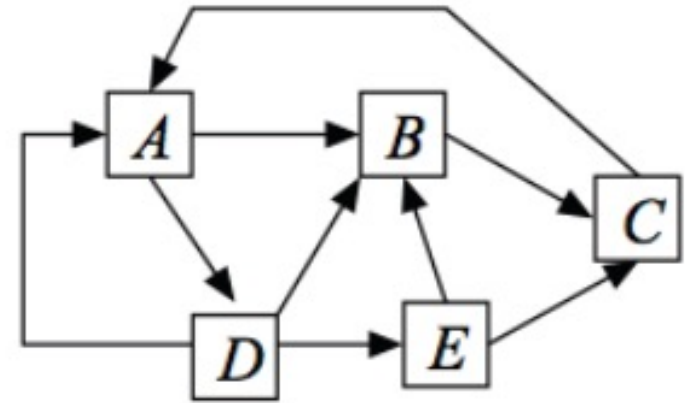
```
>> P^4  
  
ans =  
  
    0.5000    0    0.5000    0  
    0    0.5000    0    0.5000  
    0.5000    0    0.5000    0  
    0    0.5000    0    0.5000
```

```
>> P^5  
  
ans =  
  
    0    0.5000    0    0.5000  
0.5000    0    0.5000    0  
    0    0.5000    0    0.5000  
0.5000    0    0.5000    0
```

# Example 1

A MC with outgoing arrows are equally likely

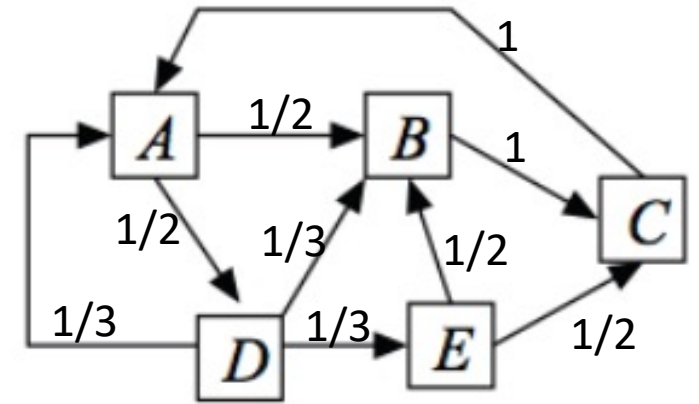
- 1) Is it irreducible?
- 2) Write transition probability
- 3) What's the most frequently visited state?



# Example 1

A MC with outgoing arrows are equally likely

- 1) Is it irreducible?
- 2) Write transition probability
- 3) What's the most frequently visited state?

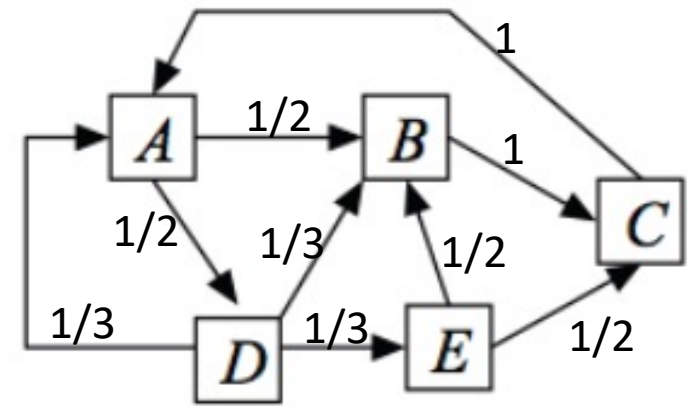




Start at A, how many steps does it take to reach E?

**Hitting time** of E starting at  $i$  is defined as

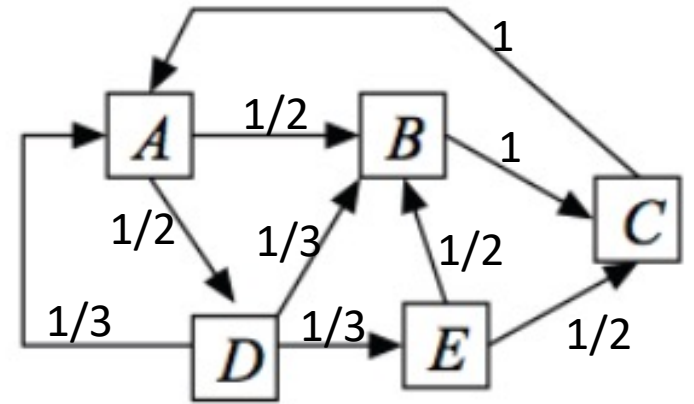
$$\beta(i) := \mathbb{E}(T_E | \mathbf{X}_0 = i) \quad \text{for } i = A, B, C, D, E$$



Goal: to calculate  $\beta(A) := \mathbb{E}(T_E | \mathbf{X}_0 = A)$

Hitting time of E starting at  $i$  is defined as

$$\beta(i) := \mathbb{E}(T_E | \mathbf{X}_0 = i) \quad \text{for } i = A, B, C, D, E$$



Goal: to calculate  $\beta(A) := \mathbb{E}(T_E | \mathbf{X}_0 = A)$

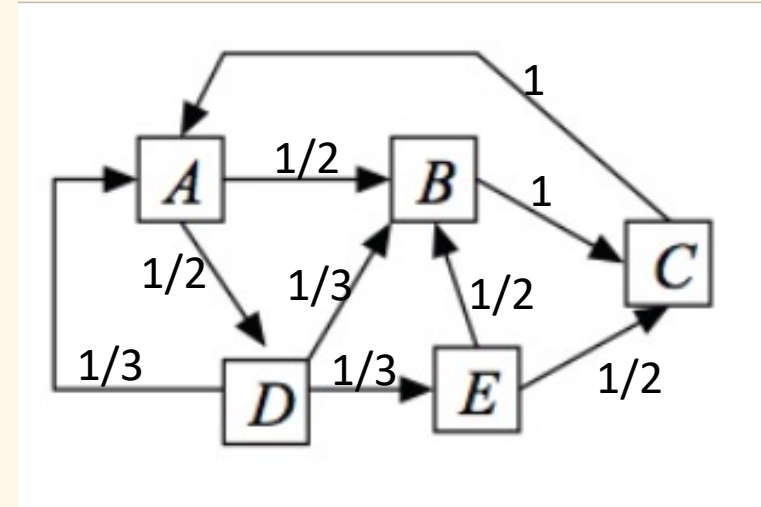
# Example 2

Flip a fair coin, how many times on average you need to flip to get two head in a row?

# Example 3

Toss a fair 6 face dice, on average, how many times we need to toss until we have the product of two number in a row is 12?

# Example 1.



What's the probability that we start at A and we visit C before we visit E

Define:

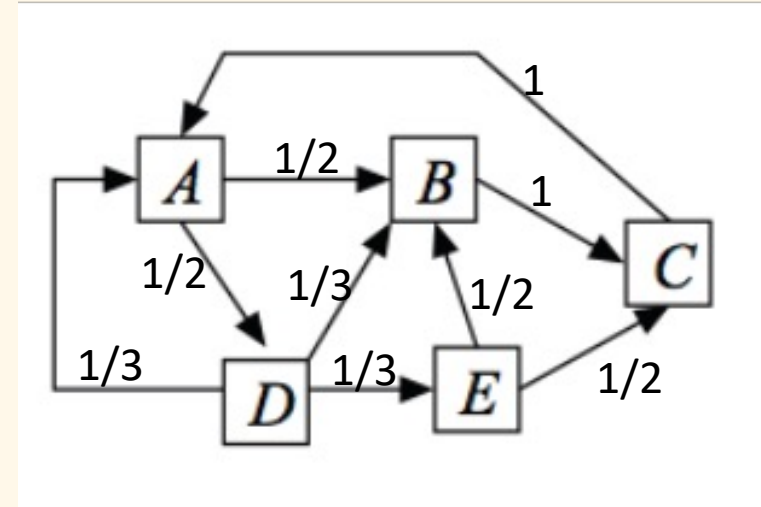
$$\alpha(i) := \mathbb{P}(T_C < T_E | \mathbf{X}_0 = i) \quad \text{for } i = A, B, C, D, E$$

Goal: to calculate  $\alpha(A) := \mathbb{P}(T_C < T_E | \mathbf{X}_0 = A)$

# Example 1

$$\alpha(i) := \mathbb{P}(T_C < T_E | \mathbf{X}_0 = i)$$

for  $i = A, B, C, D, E$



Goal: to calculate  $\alpha(A) := \mathbb{P}(T_C < T_E | \mathbf{X}_0 = A)$

# General First Step Equation (1)

For a Markov Chain on state space  $S = \{1, 2, \dots, K\}$  with transition probability  $P$ , let  $T_i$  be the hitting time of state  $i$ .

For a set  $A \subset S$  of states, let  $T_A = \min \{n \geq 0 \mid \mathbf{X}_n \in A\}$  be the hitting time of the set  $A$ .

1) We consider the mean value of  $T_A$

# General First Step Equation (2)

For a Markov Chain on state space  $S = \{1, 2, \dots, K\}$  with transition probability  $P$ , let  $T_i$  be the hitting time of state  $i$ .

For a set  $A \subset S$  of states, let  $T_A = \min \{n \geq 0 | X_n \in A\}$  be the hitting time of the set  $A$ .

2) We consider the probability of hitting set  $A$  before  $B$



# General First Step Equation (3)

3) We consider collecting an amount of  $h(i)$  every time visiting state  $i$  before visiting state A

$$Y = \sum_{n=0}^{T_A} h(\mathbf{X}_n)$$

# General First Step Equation (4)

4) We consider a discount factor  $\beta$  for moving one step

$$Z = \sum_{n=0}^{T_A} \beta^n h(\mathbf{X}_n)$$

