

Lecture 27

Markov Chains

and a Few Paradoxes.

Summary of Lecture 26

- A Markov Chain is a process that move from state to state randomly and only remembers its current state.

Ingredients:

- \mathcal{X} - Finite state space, usually $\mathcal{X} = \{1, 2, \dots, k\}$.
- π_0 - the initial distribution.
- $P(i, j)$ - Prob. to move from state i to j .

This defines a Markov Chain: sequence of r.v.s X_0, X_1, X_2, \dots

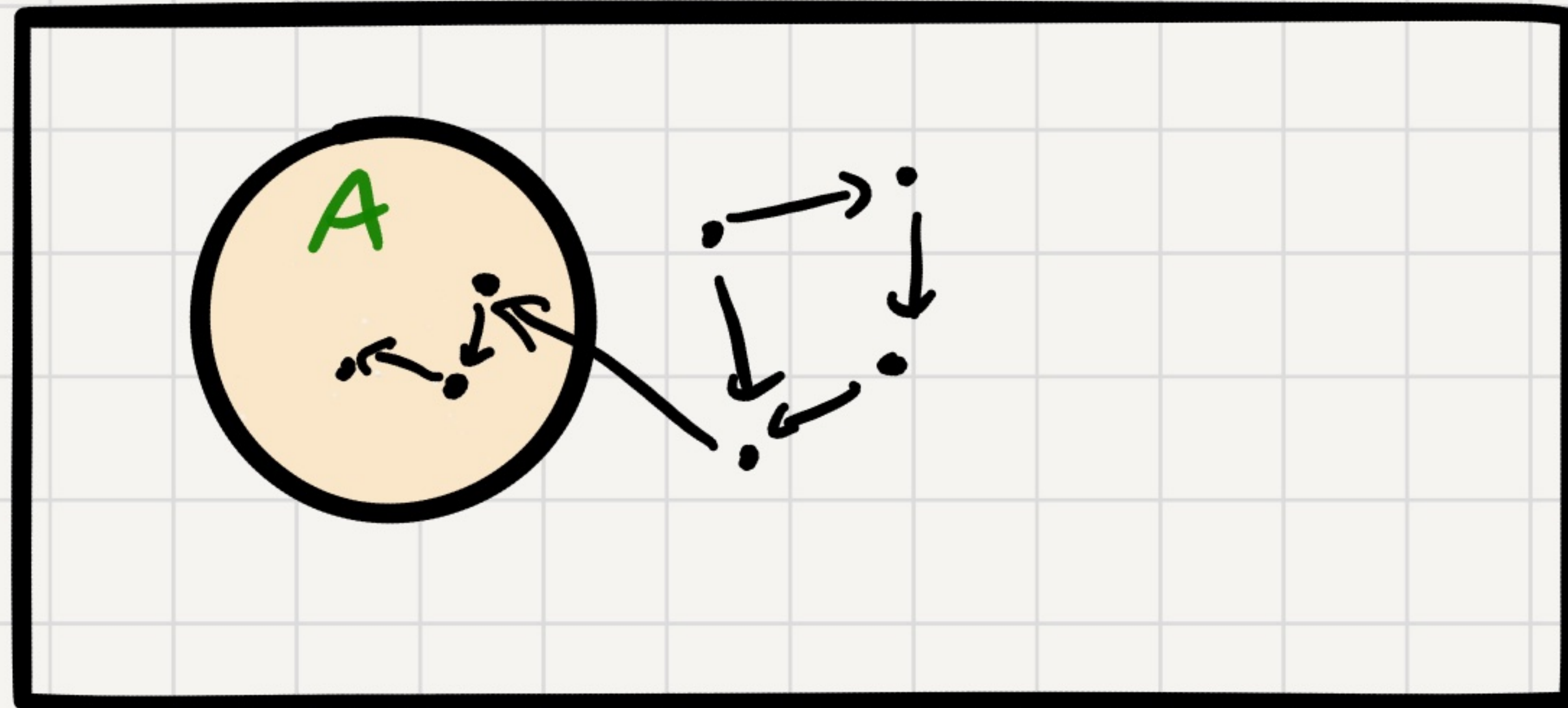
$$\Pr[X_0 = i] = \pi_0(i)$$

$$\Pr[X_{n+1} = j \mid X_n = i, X_{n-1}, \dots, X_1, X_0] = P(i, j).$$

π_n : Prob. dist of X_n .

$$\pi_n = \pi_0 \cdot P^n$$

First Step Equations



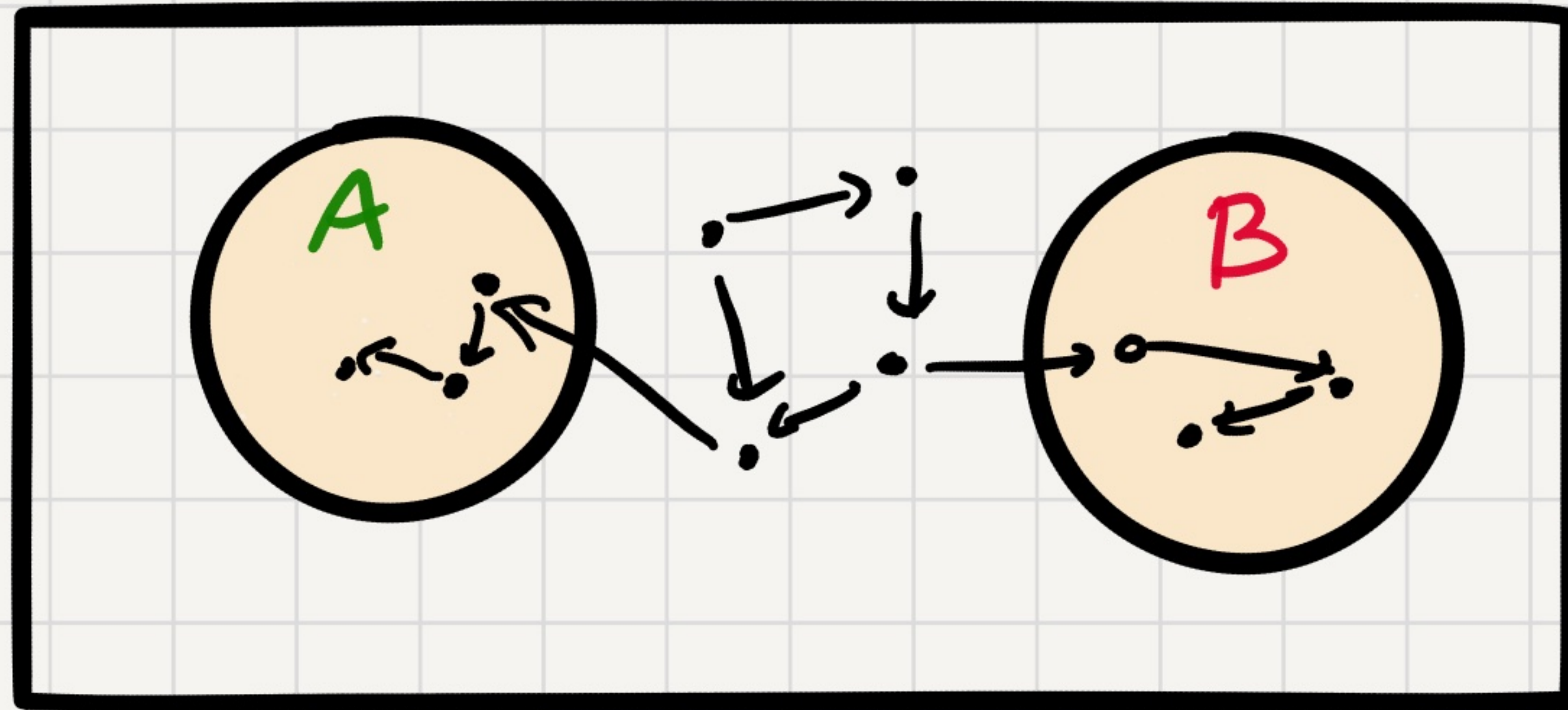
Let $\{X_n\}_{n=0}^{\infty}$ be a MC on X $A \subseteq X$

$\beta(i) =$ expected time to reach A starting from i .

$\beta(i) = 0$ for $i \in A$

$\beta(i) = 1 + \sum_j P(i,j) \beta(j)$ for $i \notin A$.

First Step Equations



Let $\{X_n\}_{n=0}^{\infty}$ be a MC on \mathcal{X} $A, B \subseteq \mathcal{X}$
 $A \cap B$ disjoint.

$\alpha(i) = \Pr[\text{reaching } A \text{ before } B, \text{ starting from } i]$

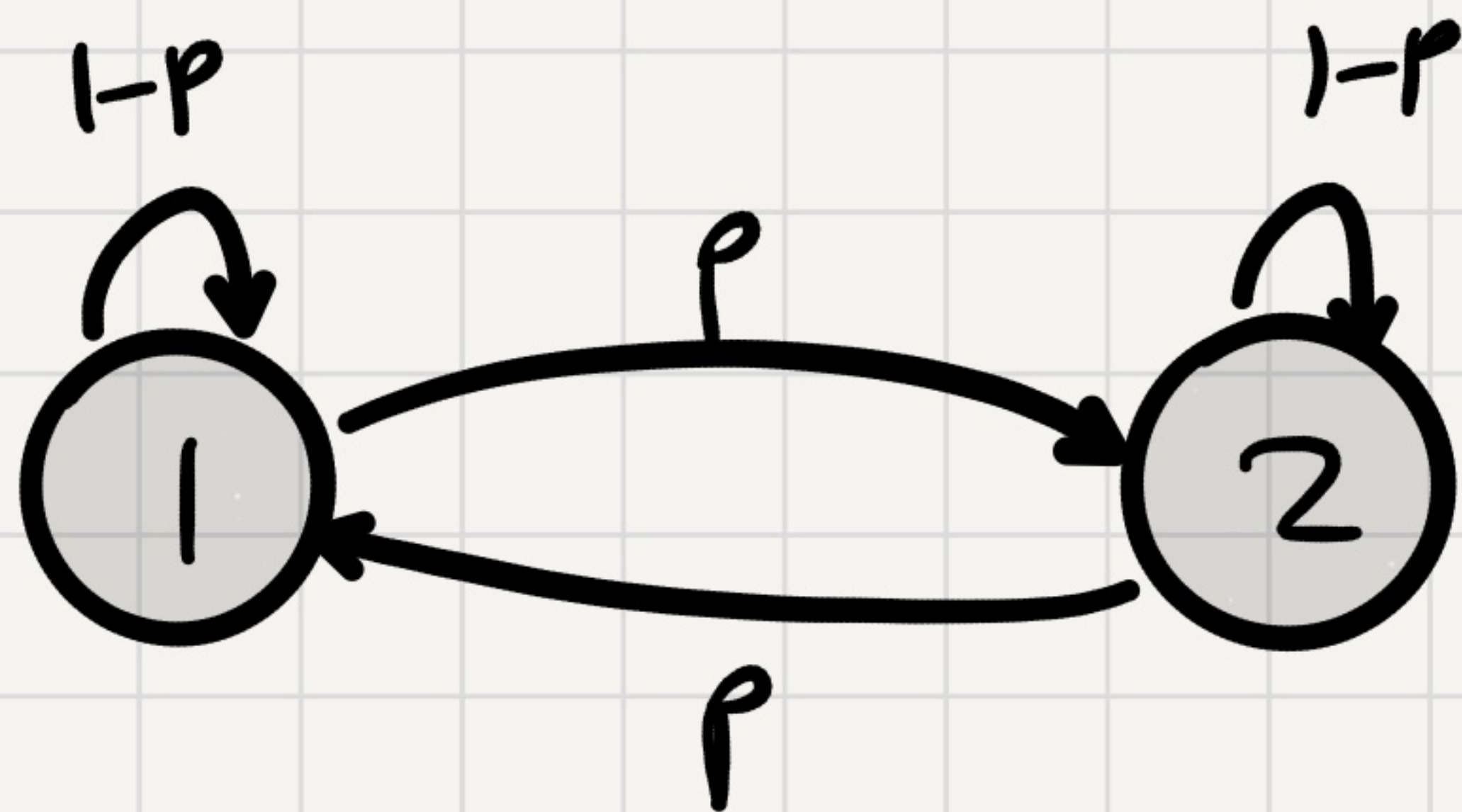
$$\left\{ \begin{array}{l} \alpha(i) = 0 \quad \text{for } i \in B \\ \alpha(i) = 1 \quad \text{for } i \in A \\ \alpha(i) = \sum_j P(i,j) \cdot \alpha(j) \quad \text{for } i \notin A \cup B. \end{array} \right.$$

Definition:

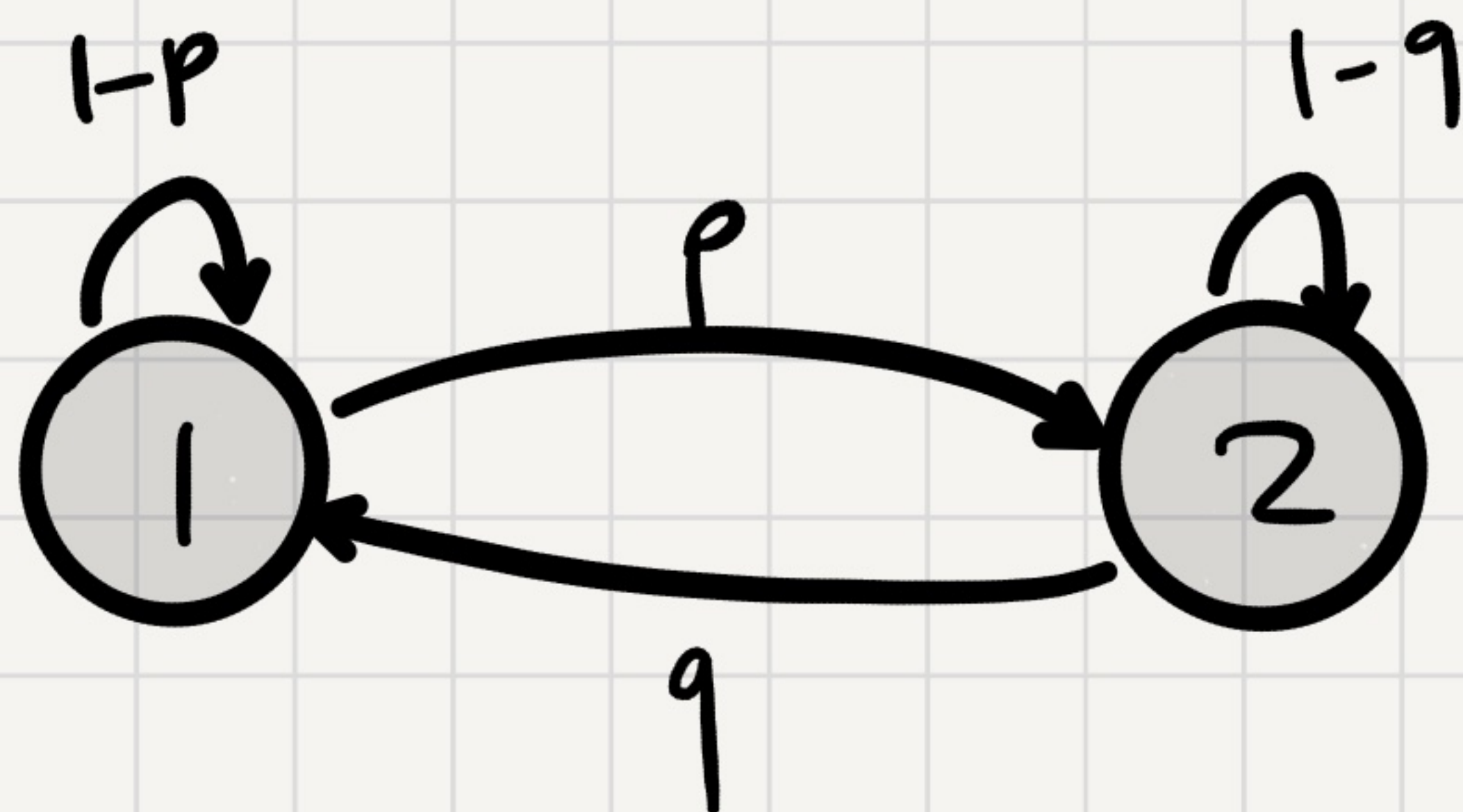
A distribution π over \mathcal{X} is stationary
(aka invariant) if $\pi = \pi P$.

If π_0 is stationary then $\forall n \quad \pi_n = \pi_0$.

Stationary Distribution-Example



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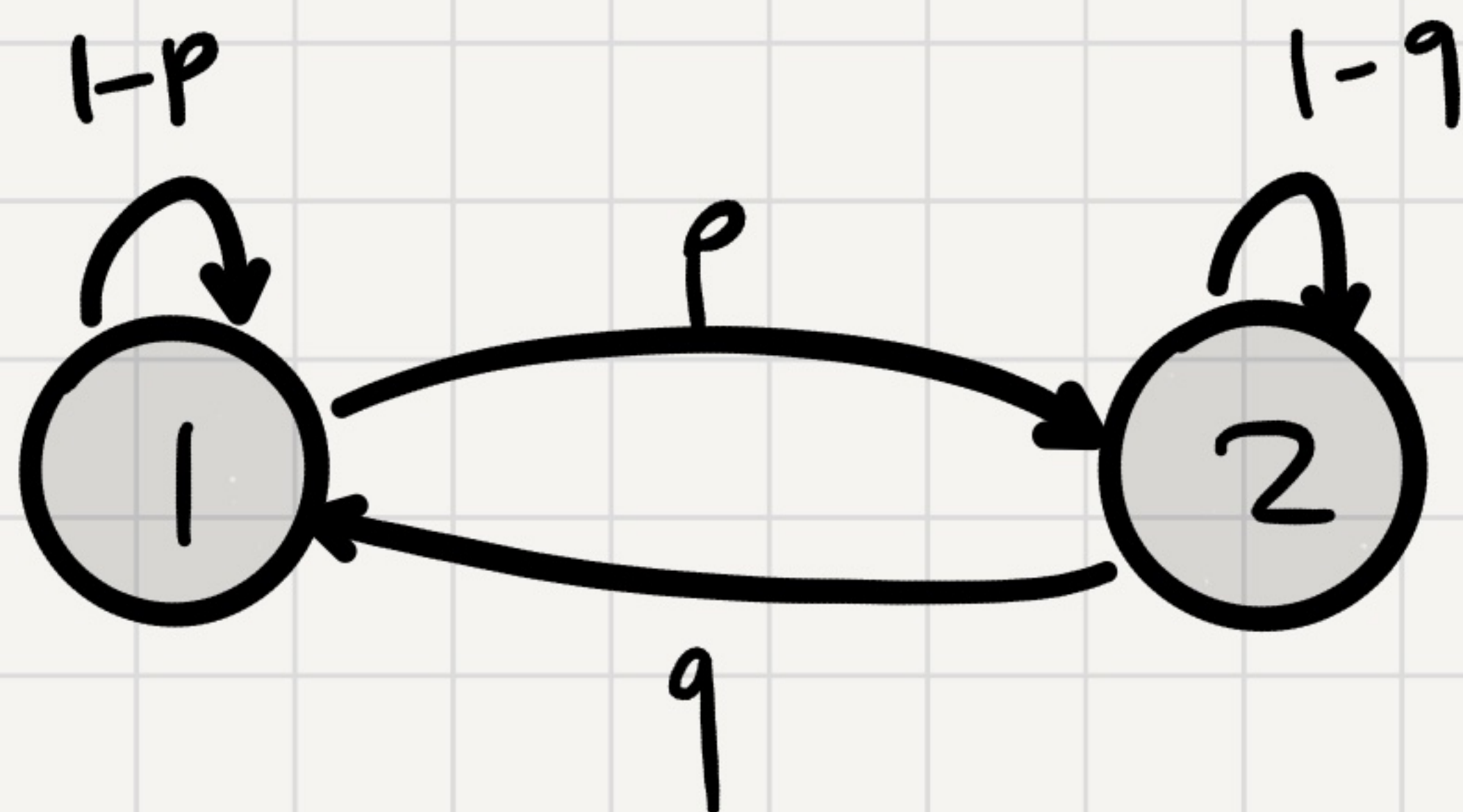


π is stationary iff $(\pi(1) \ \pi(2)) = (\pi(1) \ \pi(2)) \cdot \begin{pmatrix} 1-p & p \\ q & 1-q \end{pmatrix}$

$$\pi(1) = \pi(1) \cdot (1-p) + \pi(2) \cdot q$$

$$\pi(2) = \pi(1) \cdot p + \pi(2) \cdot (1-q)$$

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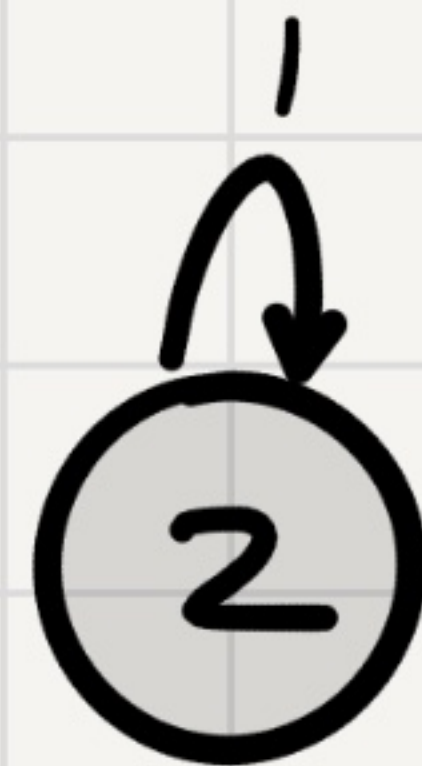
$$\pi(1) = \pi(1) \cdot (1-p) + \pi(2) \cdot q \quad \Leftrightarrow \quad \pi(1) \cdot p = \pi(2) \cdot q$$

$$\pi(2) = \pi(1) \cdot p + \pi(2) \cdot (1-q) \quad \Leftrightarrow \quad \pi(1) \cdot p = \pi(2) \cdot q$$

$$\pi(1) + \pi(2) = 1$$

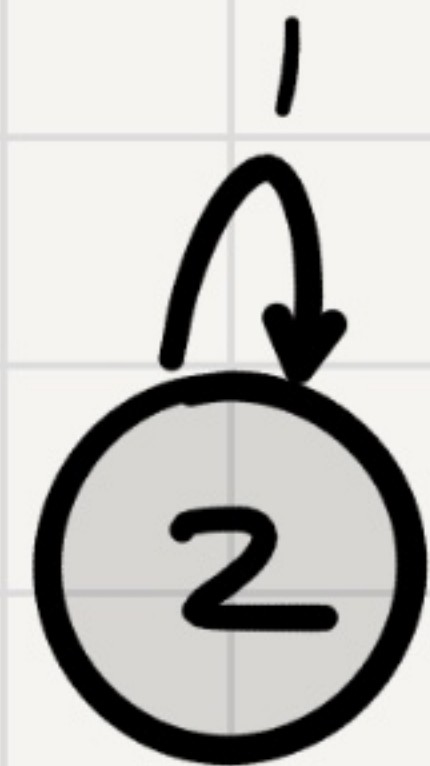
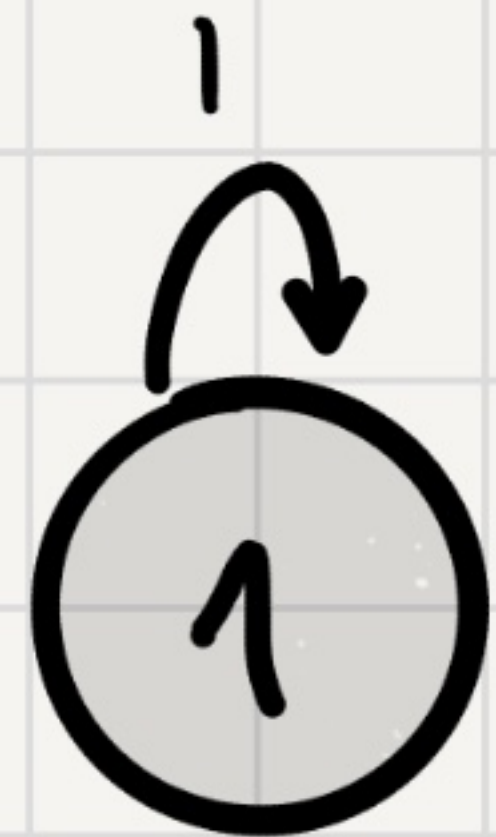
Solution: $\pi = \left[\frac{q}{p+q}, \frac{p}{p+q} \right]$

Stationary Distributions - Example 2



Which distributions are stationary?

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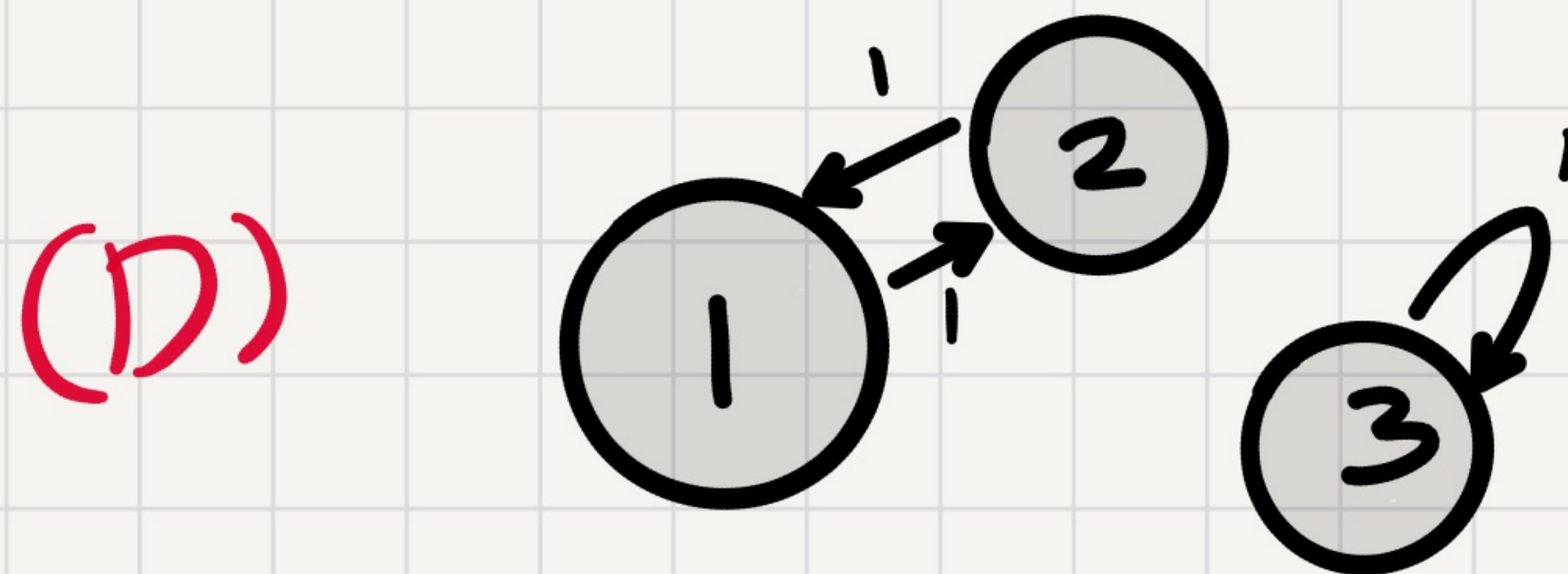
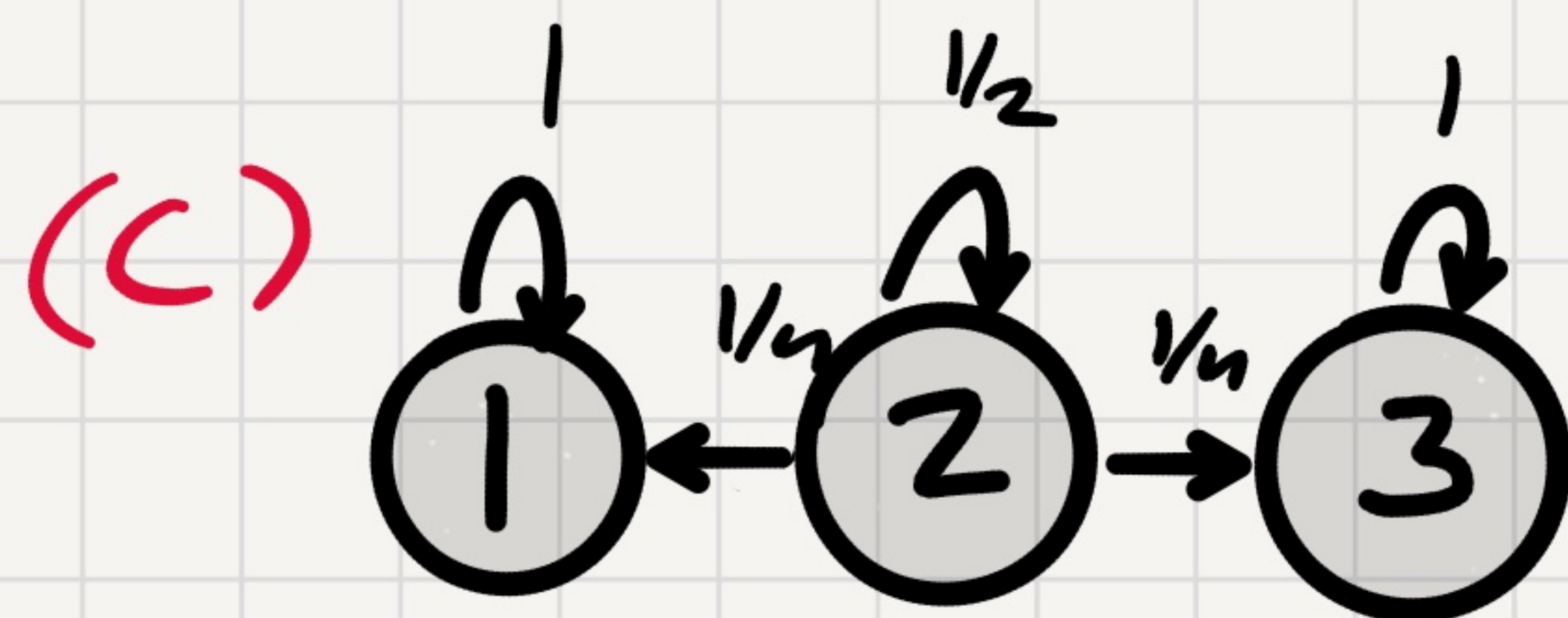
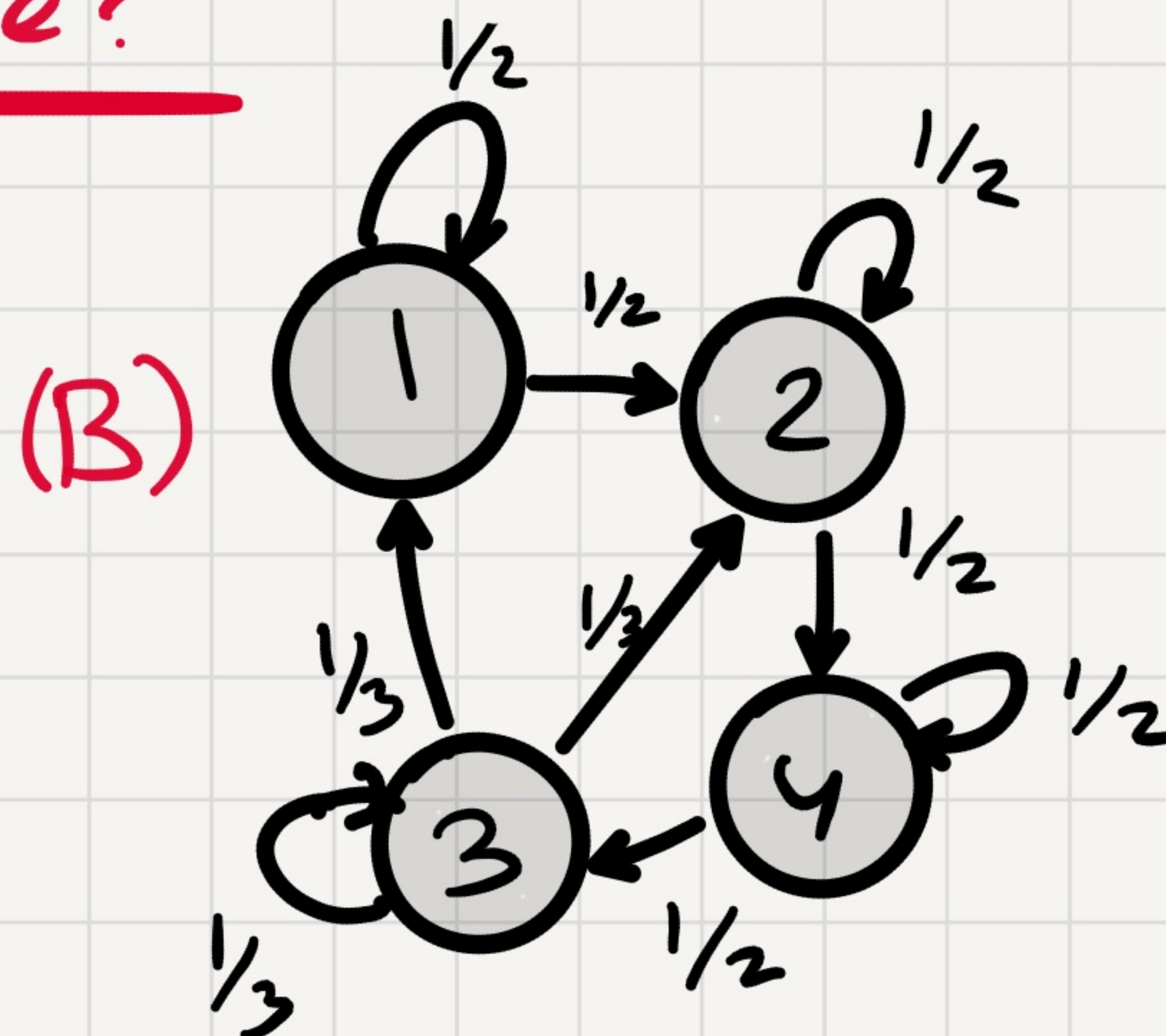
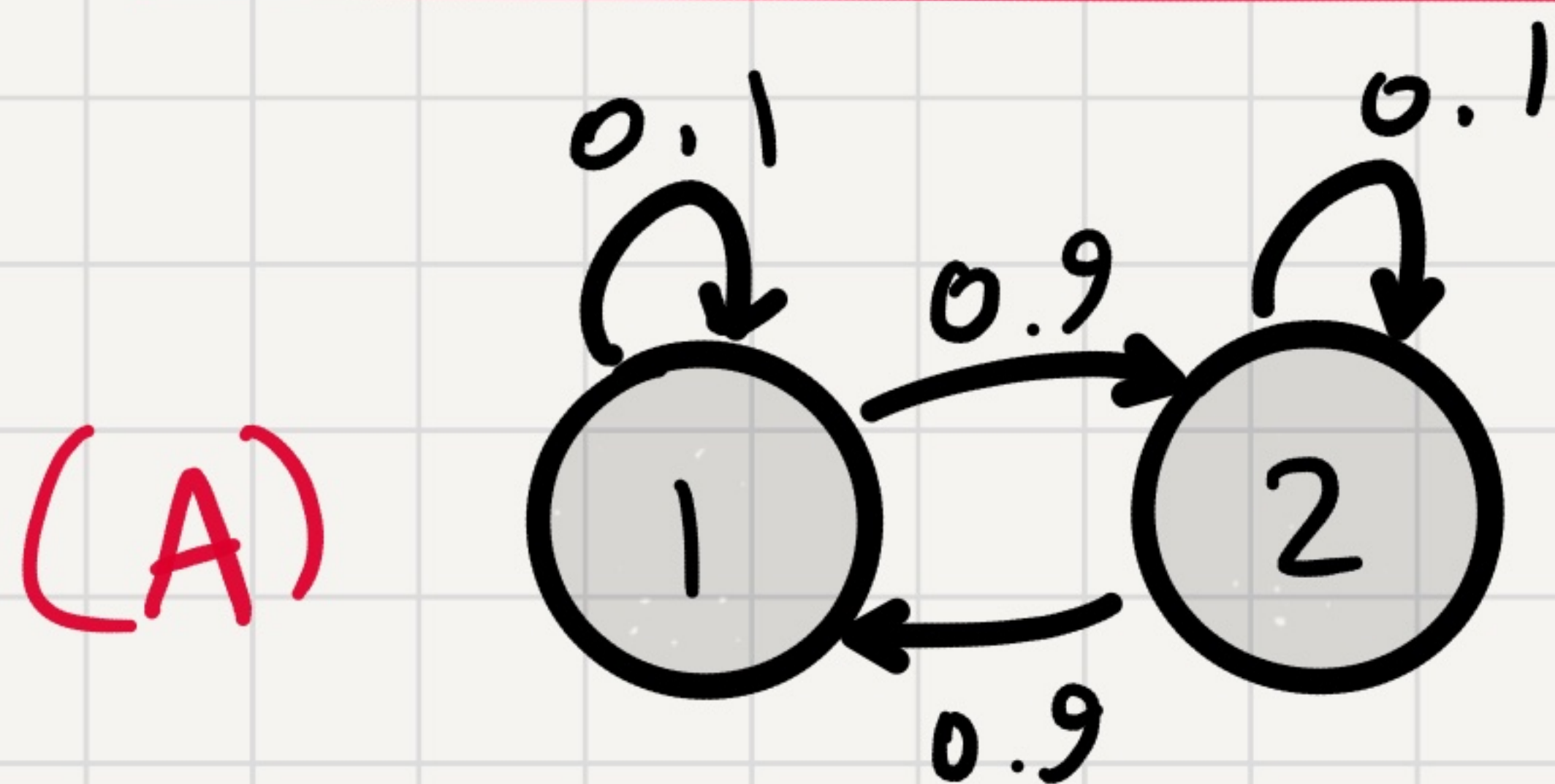
all of them.

$$\forall \pi \quad \pi = \pi \cdot \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

Irreducible Markov Chains

A MC is irreducible if you can go from every state i to every state j (possibly in multiple steps).

Which MC are irreducible?



Theorem:

Any finite irreducible MC has one and only one stationary distribution.

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Theorem 2: (Long Term Fraction of Time in States)

If $(X_n)_{n=0}^{\infty}$ is an irreducible MC on $\{1, \dots, k\}$

with stationary distribution π .

Then, for any start dist. π_0 , for all $i \in \{1, \dots, k\}$

$$\frac{1}{n} \sum_{m=0}^{n-1} \mathbb{1}_{\{X_m = i\}} \xrightarrow{n \rightarrow \infty} \pi(i).$$

Intuition: Start at a dist. π_0 and suppose

the limits exist. Denote by

$$f(i) = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{m=0}^{n-1} \mathbb{1}_{\{x_m = i\}}.$$

What's the frac. of times we visit i ?



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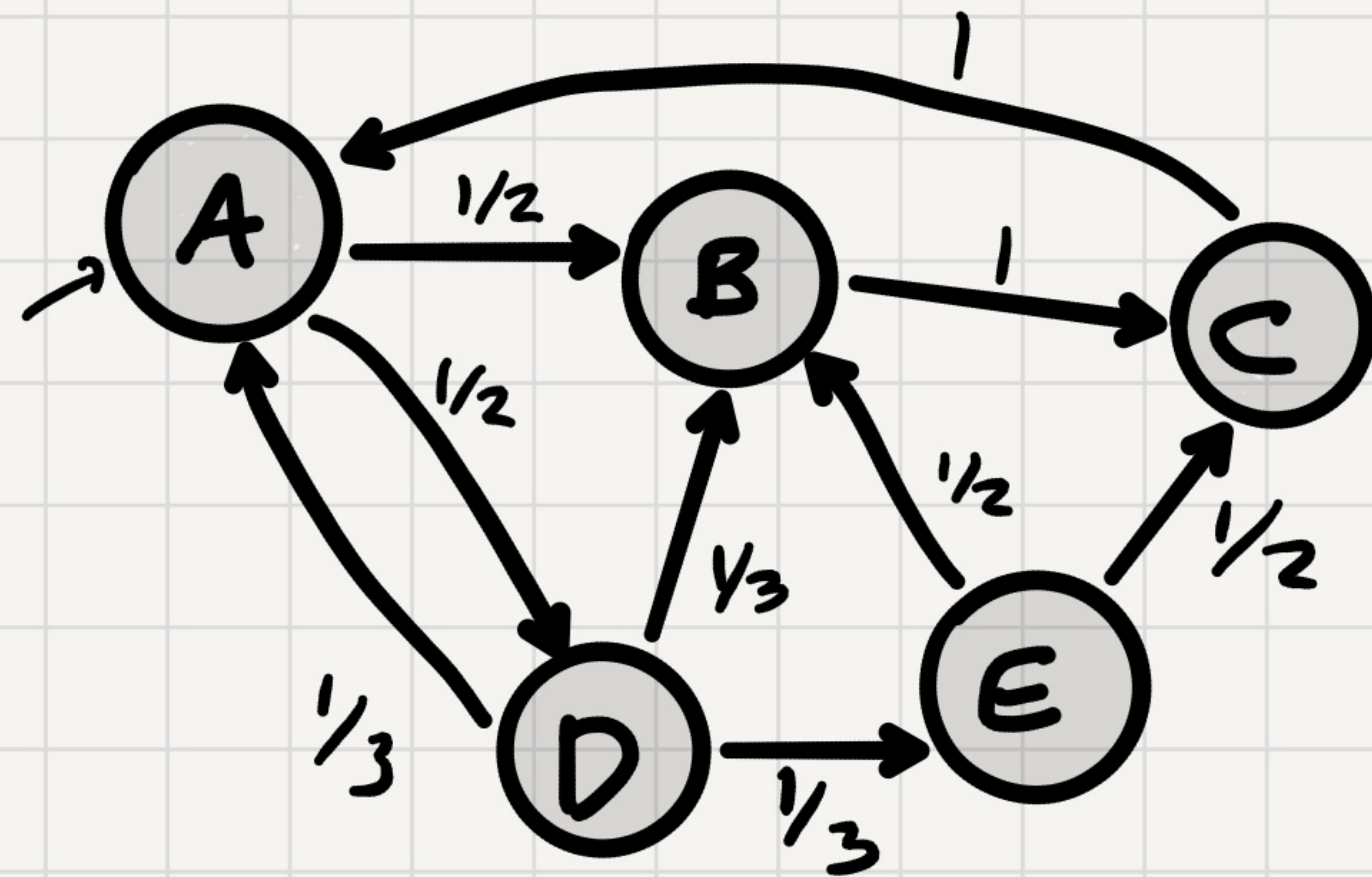
What's the frac. of times we visit i and then move to j ? 

$$\text{Frac. of times we're at } j = \sum_i f(i) \cdot P(i, j)$$

$\forall j$ $f(j) = \sum_i f(i) \cdot P(i, j)$. Hence, f is a stationary dist.

Let's see a Simulation:

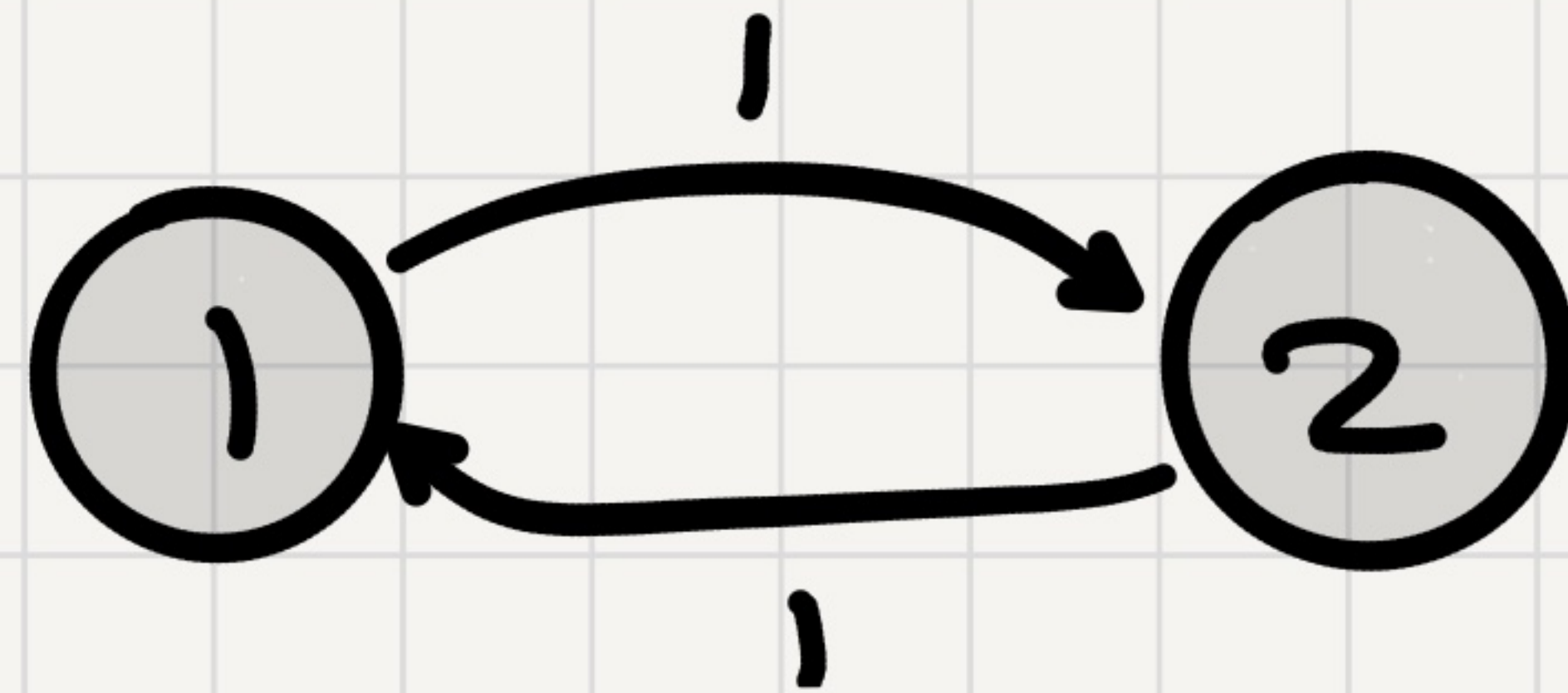
Recall this MC from lecture 26.



We'll run it for many steps and count how many times we've been in each step.

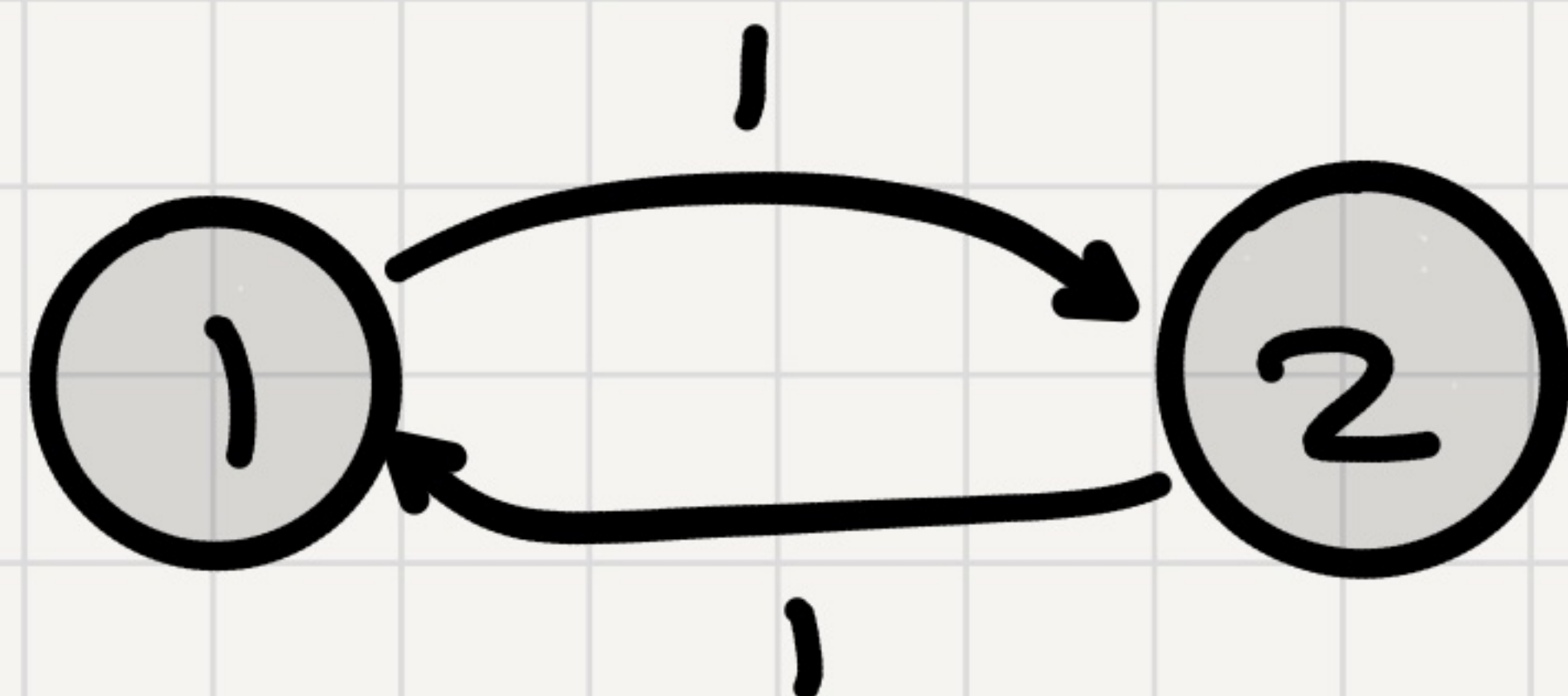
Converges to the Stationary Distribution

Example:



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Example:



The MC is irreducible.

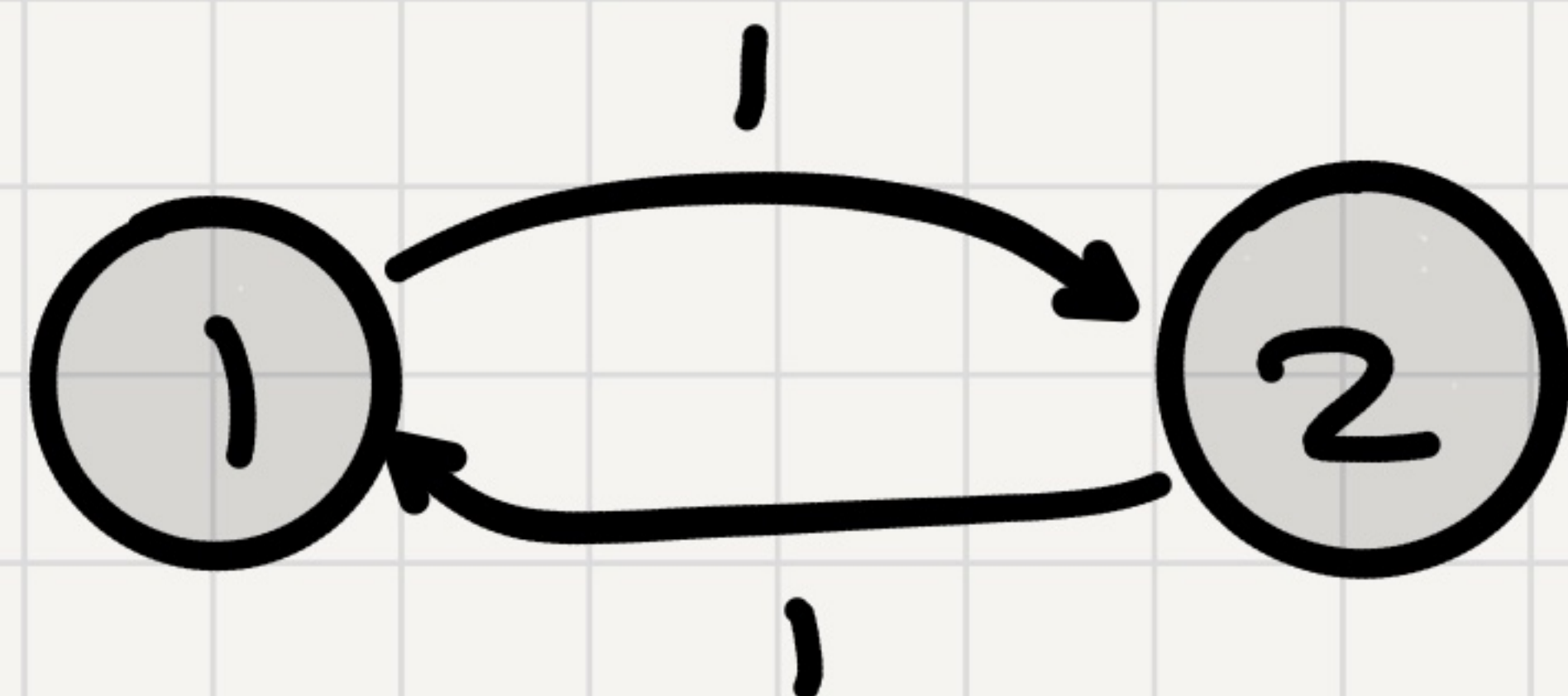
Its stationary dist satisfies

$$(\pi(1) \quad \pi(2)) = (\pi(1) \quad \pi(2)) \cdot \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$$

$$\Rightarrow \pi(1) = \pi(2) = 1/2.$$

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But starting from 1: $\pi_0 = (1 \ 0)$
 $\pi_1 = (0 \ 1)$
 $\pi_2 = (1 \ 0) \dots$

$$\pi_{2m} = (1 \ 0)$$
$$\pi_{2m+1} = (0 \ 1)$$

Periodicity

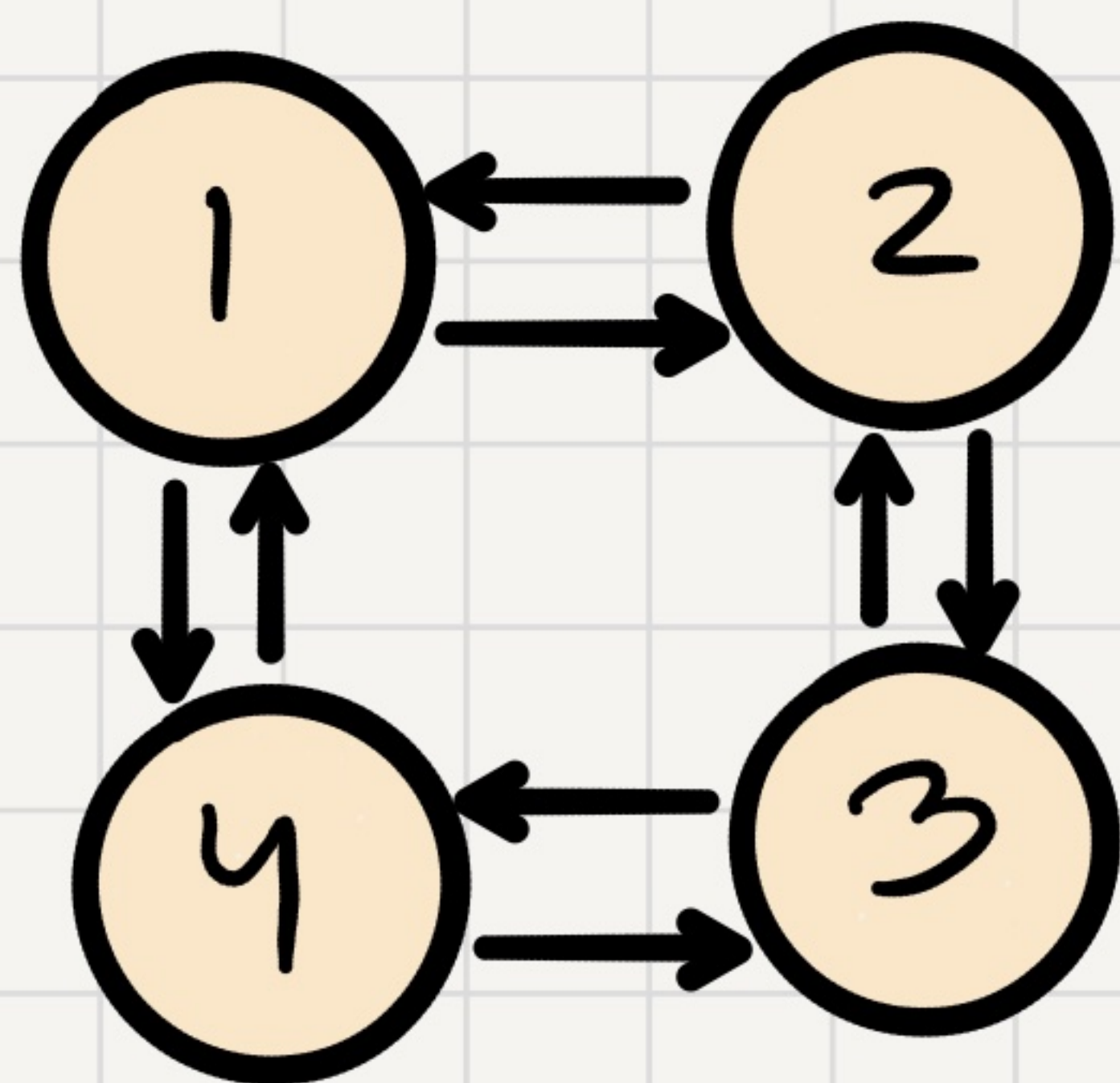
Def'n: The periodicity of a Markov Chain with transition matrix P is the gcd of lengths of all closed walks in the chain

$$\text{gcd}(n > 0 \mid \exists i \text{ s.t. } P^n(i, i) > 0)$$

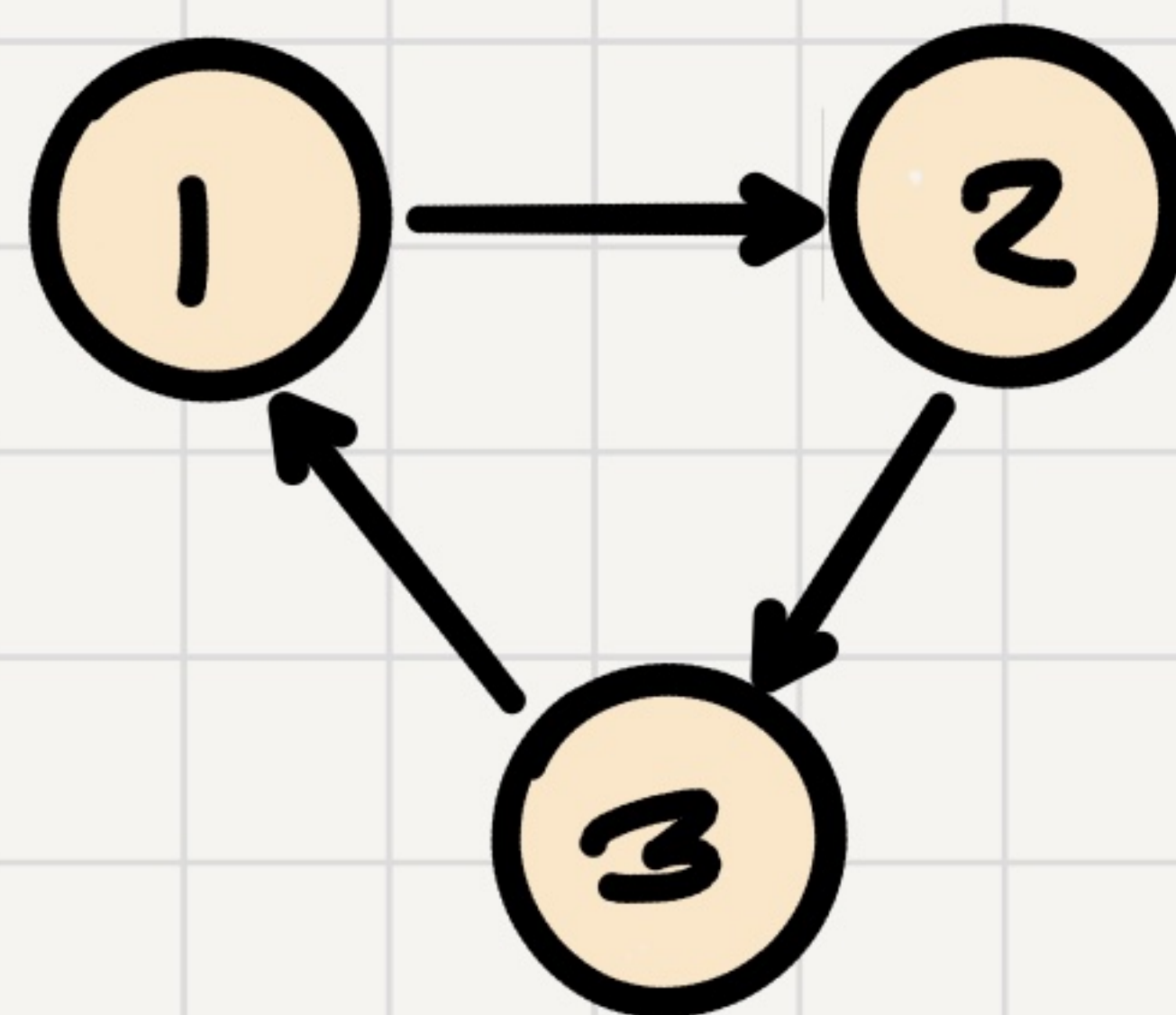
A Markov chain is aperiodic if this $\text{gcd} = 1$.

Which Markov Chains are Aperiodic?

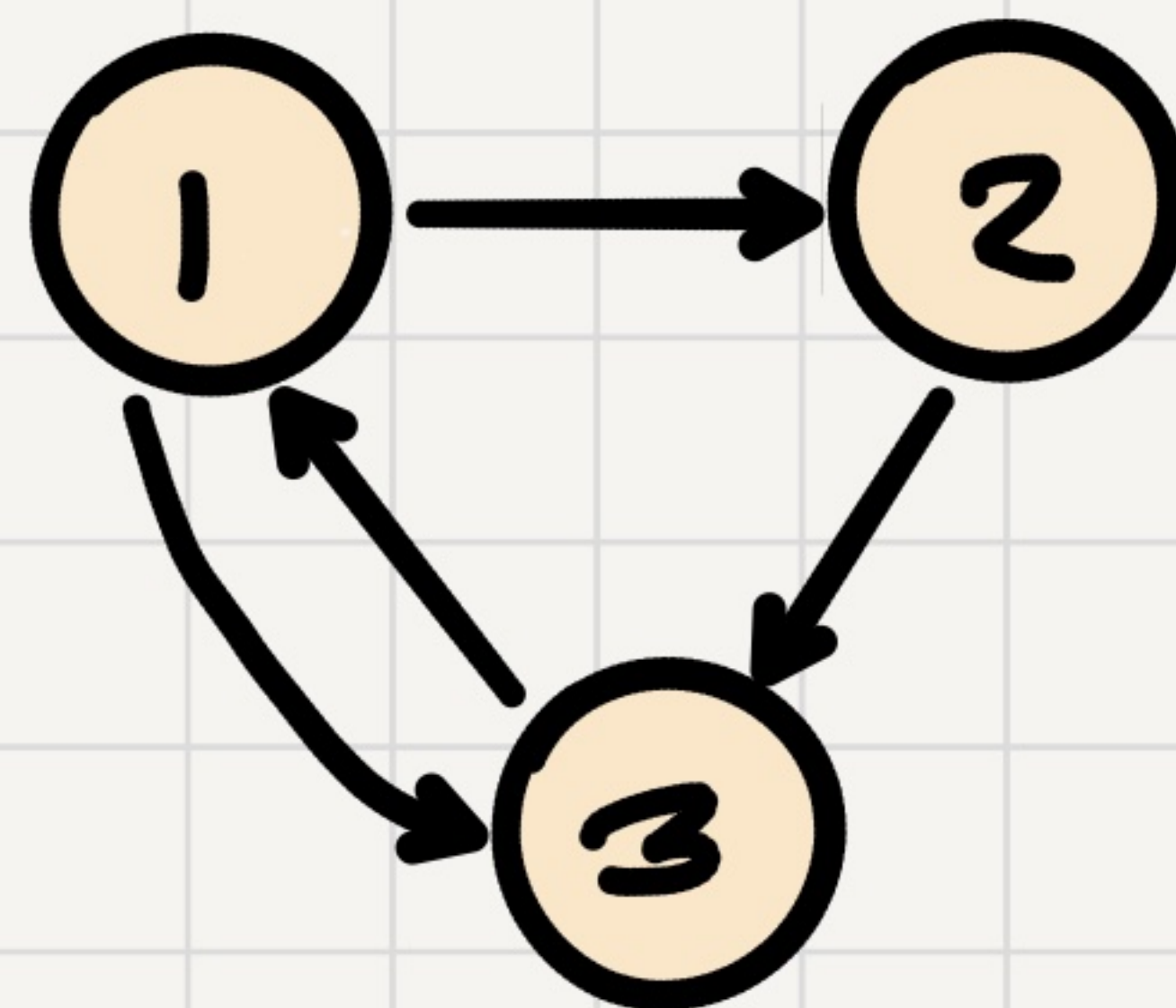
A.



B.



C.



Theorem: Let X_n be an irreducible
a periodic Markov chain with stationary dist π .

Then, no matter what's the starting dist. π_0

$$\forall i: \pi_n(i) \xrightarrow{n \rightarrow \infty} \pi(i).$$

Some Paradoxes

St. Petersburg Paradox

A casino is offering you to play the following game:

- Start with a stake of z \$.
- At each point flip a fair coin
 - H: double the stake.
 - T: stop and give stake to player.

How much are you willing to pay to play this game?

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What's the expected winning stake?

X r.v. capturing winning stake

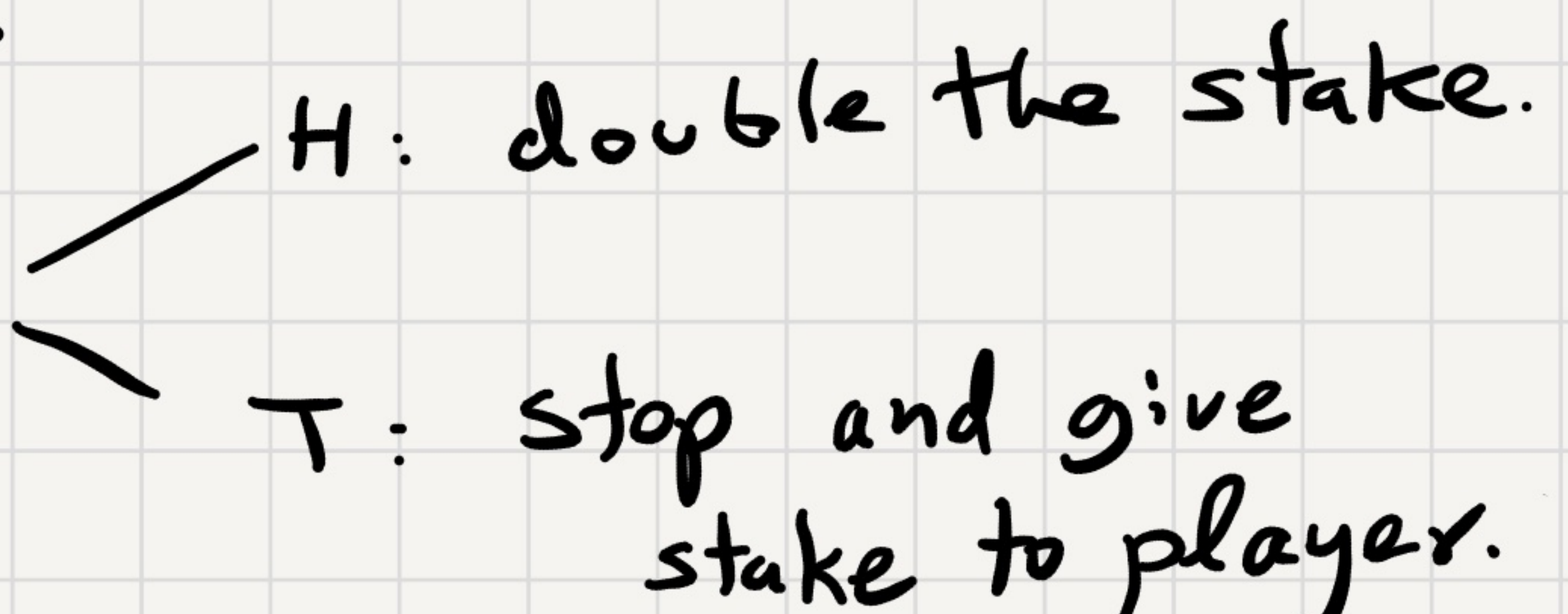
$$P_r[X=2] = 1/2 \quad P_r[X=4] = 1/4 \quad \dots \quad P_r[X=2^k] = \frac{1}{2^k}.$$

$$E[X] = \sum_{i=1}^{\infty} P_r[X=2^i] \cdot 2^i = \sum_{i=1}^{\infty} \frac{1}{2^i} \cdot 2^i = \sum_{i=1}^{\infty} 1 = \infty.$$

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$$\begin{aligned} E[X] &= 2 \cdot \frac{1}{2} + 4 \cdot \frac{1}{4} + \dots + 2^{k-1} \cdot \frac{1}{2^{k-1}} + 2^k \cdot \frac{1}{2^{k-1}} \\ &= k+1 = \log_2 n + 1 \end{aligned}$$

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What happens if you have a limited budget, say $1023\$$?

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Eventually a H will be flipped and you will win overall $1\$$.

What happens if you have a limited budget, say $1023\$$:

You'll lose $1023\$$ with prob. $\frac{1}{1024}$
and win $1\$$ with prob. $1 - \frac{1}{1024}$

On expectation, you'll even out.

Confusing Statistics:

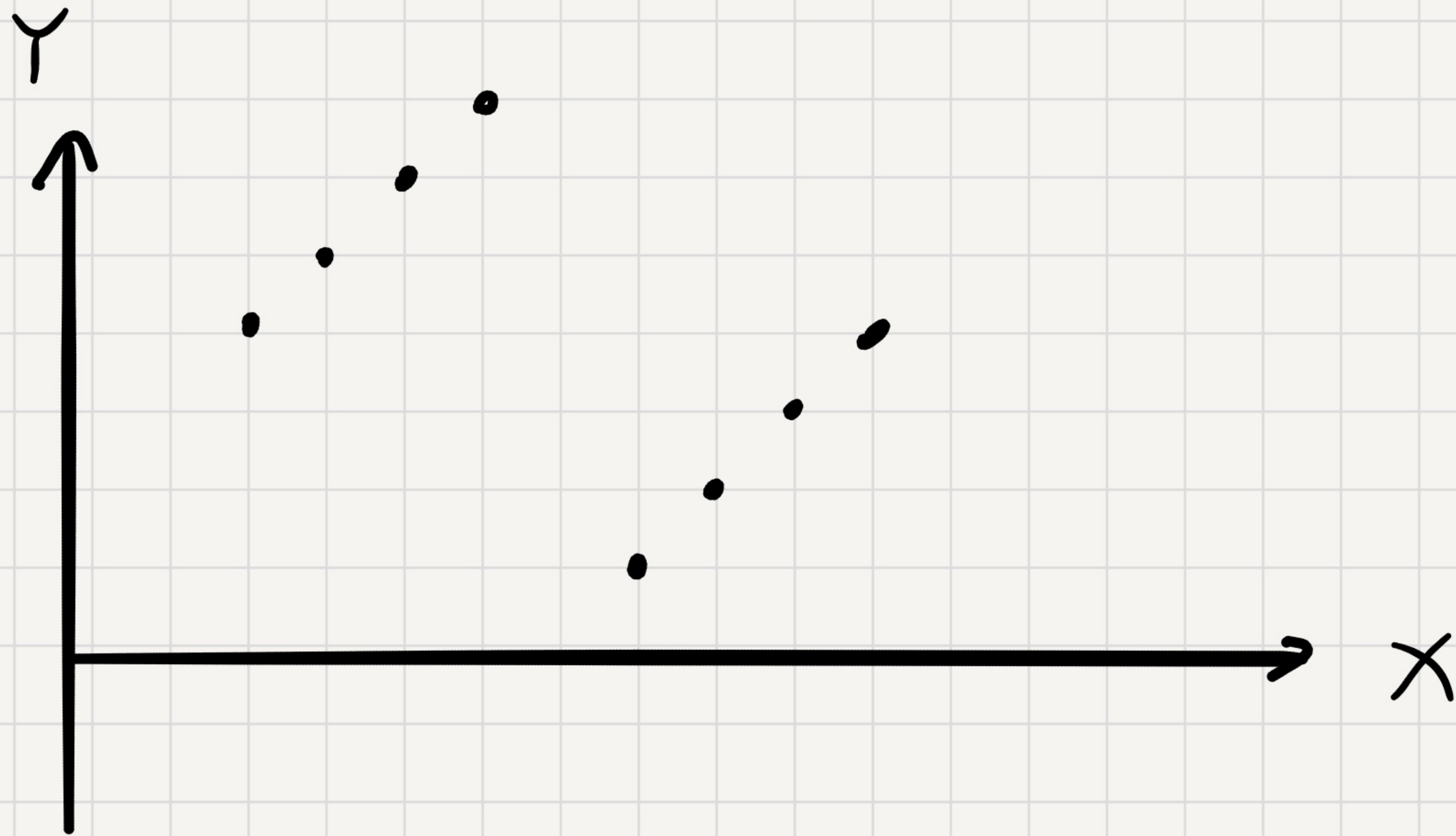
Simpson's Paradox

Results from real-life medical study:

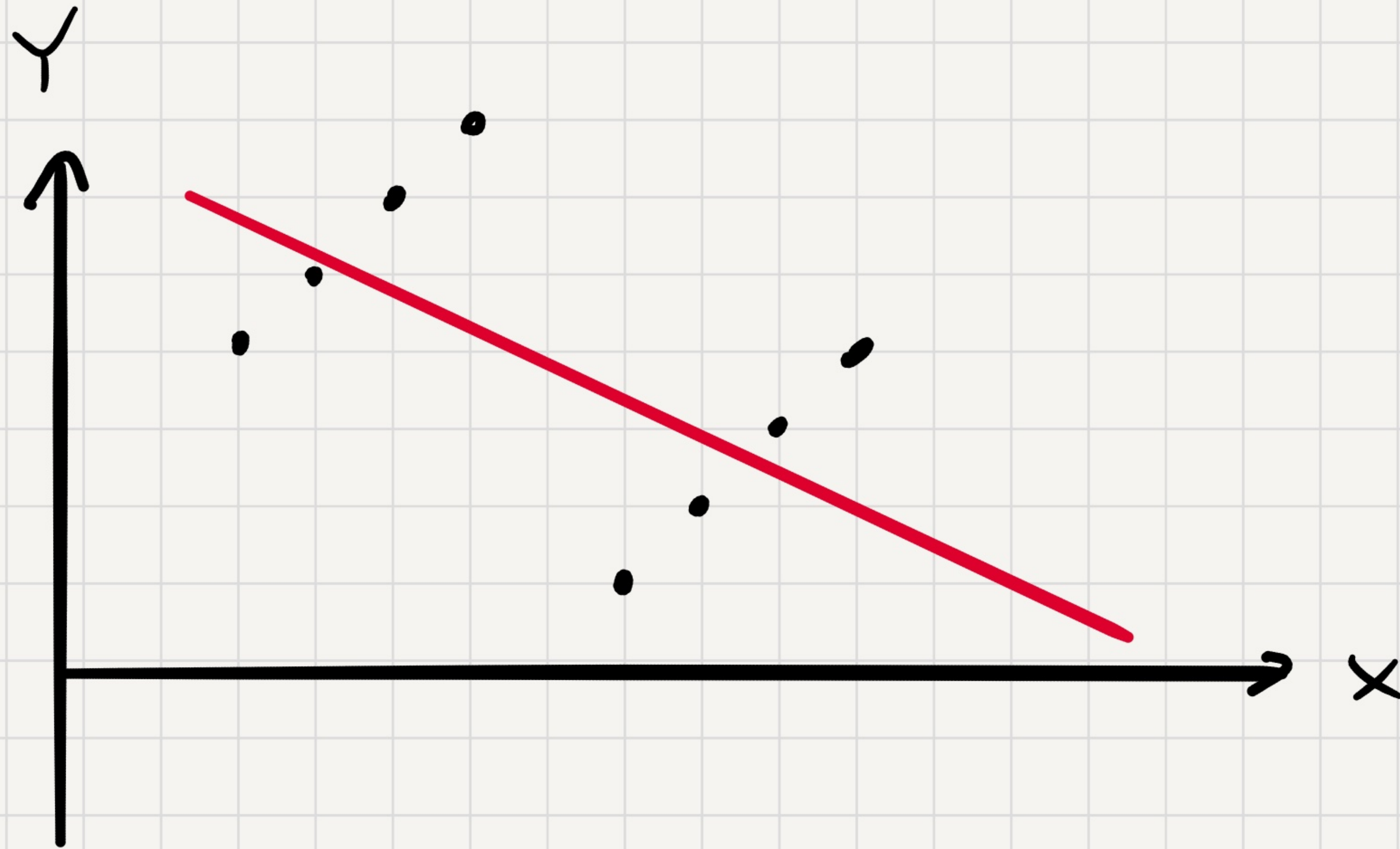
	Treatment A	Treatment B
Small Stones	Group 1 93% (81/87)	Group 2 87% (234/270)
Large Stones	Group 3 73% (192/263)	Group 4 69% (55/80)
Both	78% (273/350)	83% (289/350)

Which treatment is better?

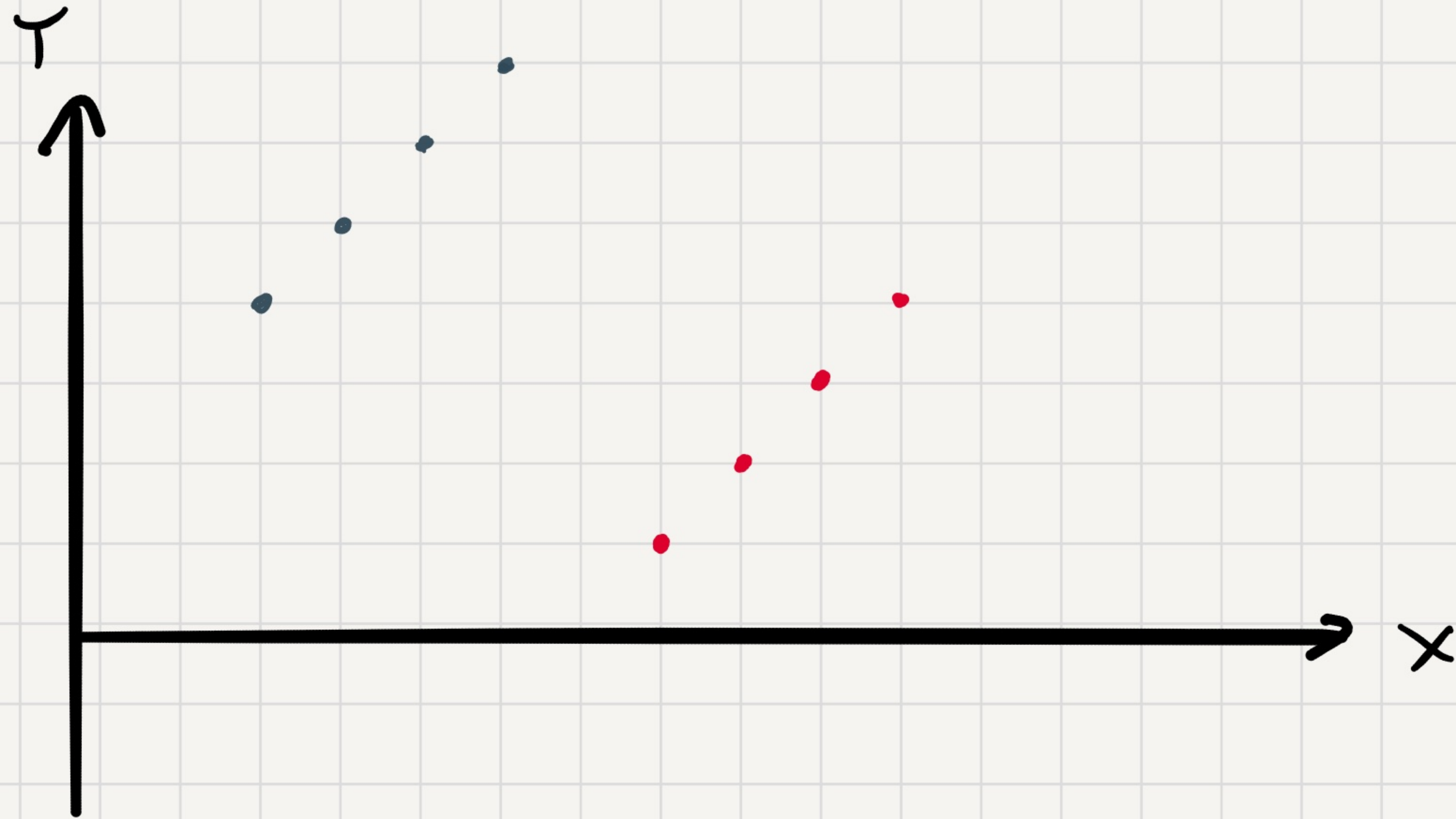
What's the best linear predictor of Y given X ?



What's the best linear predictor of Y given X ?



But if the data is coming from a mixture of two distributions **red** and **blue**:



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Thank you
and good luck in
the final!