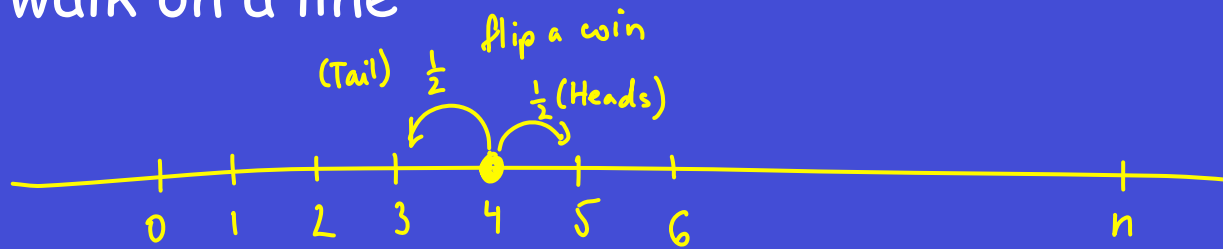


Introduction to Markov chains

Examples of Markov chains:

- Random walk on a line

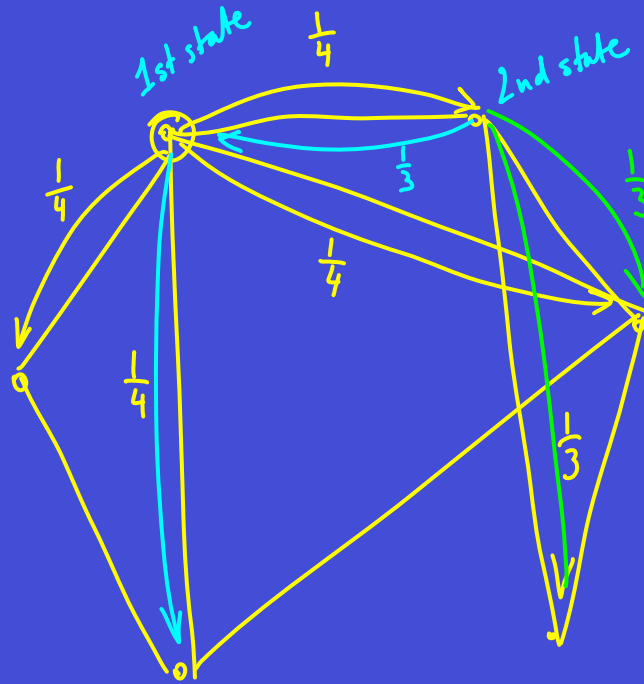


HTTHTHTTT

Introduction to Markov chains

Examples of Markov chains:

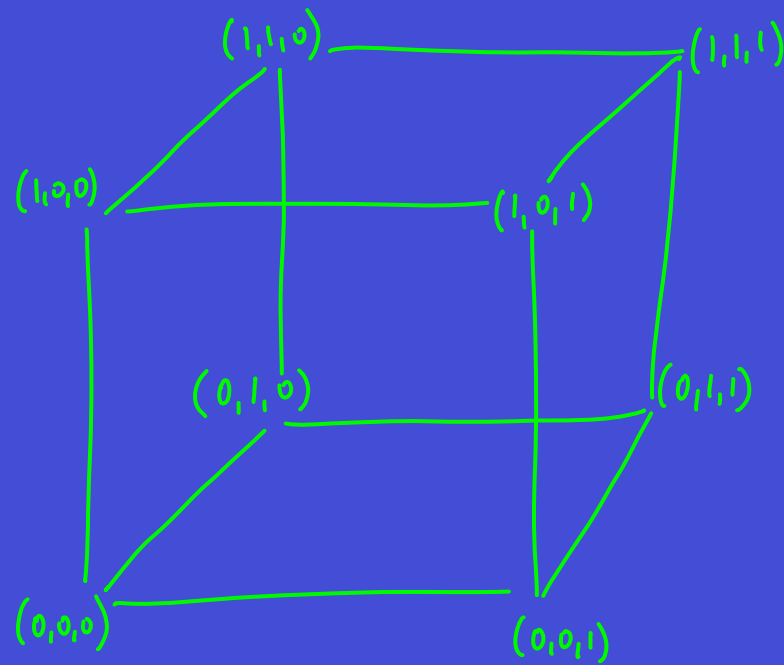
- Random walk on a graph



Introduction to Markov chains

Examples of Markov chains:

- Random walk on a hypercube



States: all bin. vectors
of length n

Move:

- choose a random position
from 1 to $n \rightarrow$ let be i
- flip i -th bit

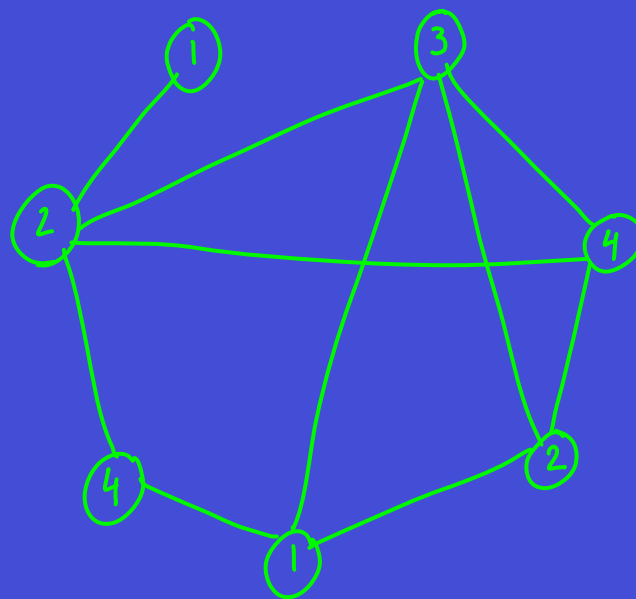
Symmetric: $\text{prob. of moving from state } u \text{ to state } v$
 $= \text{prob. of moving from state } v \text{ to state } u$

Introduction to Markov chains

Examples of Markov chains:

- Markov chain on graph colorings

states:
valid colorings



colors is: 4

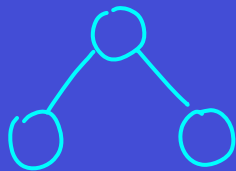
MC step:

- randomly choose a vertex v
- randomly choose a color c
- recolor the vertex v with color c
(if valid)

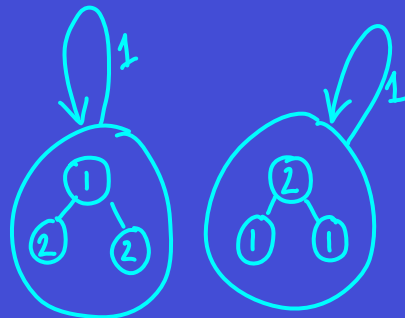
Introduction to Markov chains

Examples of Markov chains:

- Markov chain on graph colorings



# colors :	1	# states :	0
	: 2		: 2
	: 3		

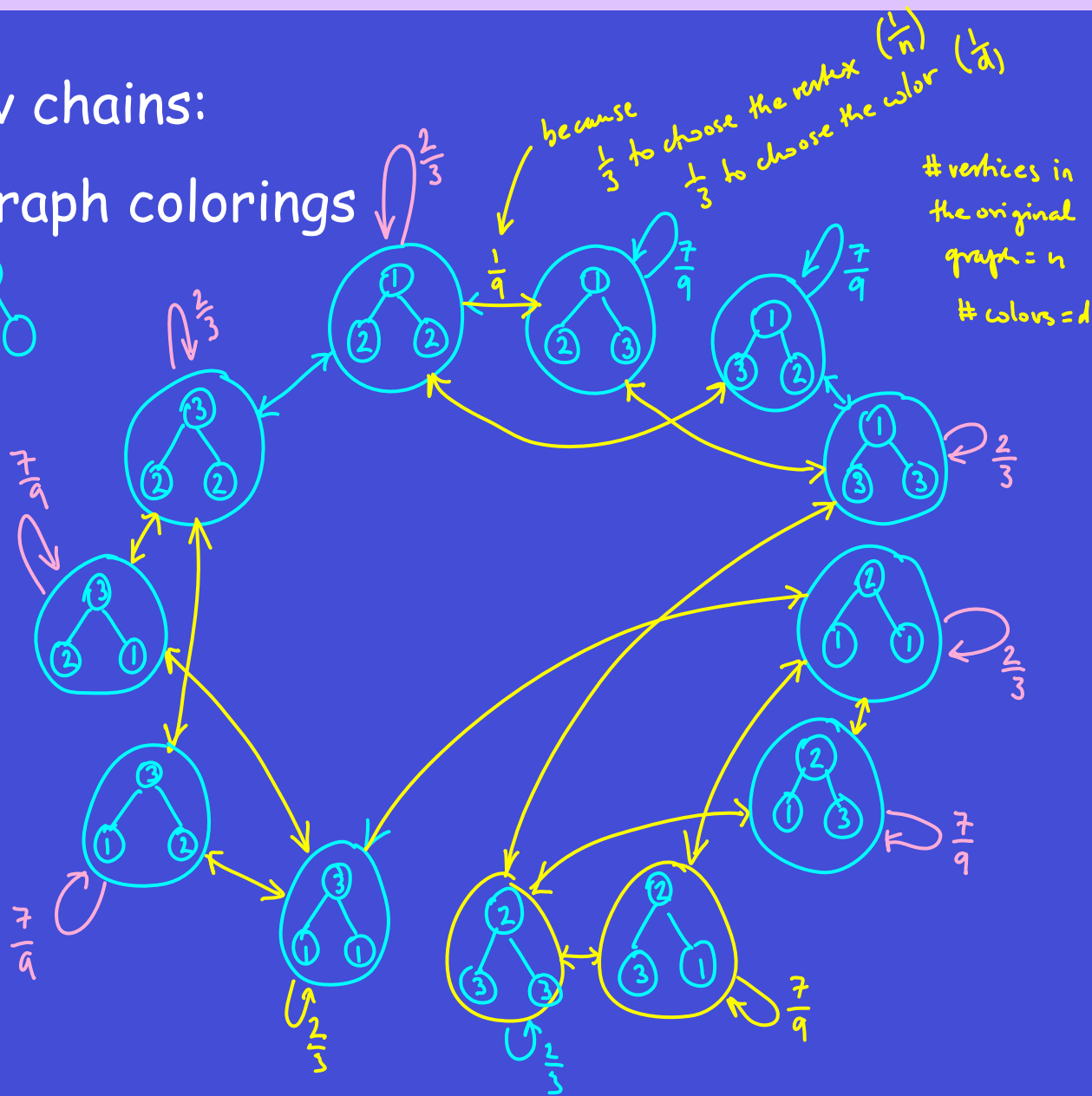
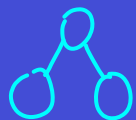


Introduction to Markov chains

Examples of Markov chains:

- Markov chain on graph colorings

colors: 3



all yellow arcs
probability $\frac{1}{9}$

if input graph has n vertices,
and we have d colors,
how many possible colorings?

up to d^n colorings
exponential
in n

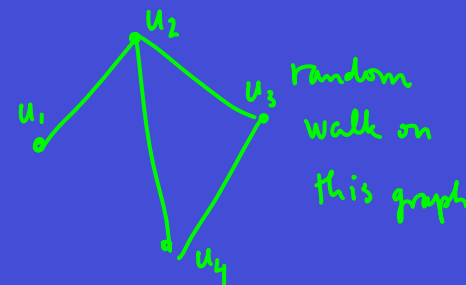
Introduction to Markov chains

Examples of Markov chains:

- Markov chain on matchings of a graph

Definition of Markov chains

- State space Ω \rightarrow the set of states
- Transition matrix P , of size $|\Omega| \times |\Omega|$



- $P(x,y)$ is the probability of getting from state x to state y

- $\sum_{y \in \Omega} P(x,y) = 1$ a prob. distribution
(P is a stochastic matrix)

if start in state u_1 , then prob. of being in state u_2 is $\frac{1}{3}$,
 u_j is 0, $j \in \{2,3,4\}$

initial distribution in this case is $(1, 0, 0, 0)$

after 1 step of the MC $(0, 1, 0, 0)$

2nd $(\frac{1}{3}, 0, \frac{1}{3}, \frac{1}{3})$

3rd $(0, \frac{1}{3} + \frac{1}{6} + \frac{1}{6}, \frac{1}{6}, \frac{1}{6})$
 $= (0, \frac{2}{3}, \frac{1}{6}, \frac{1}{6})$

$$\Omega = \{u_1, u_2, u_3, u_4\}$$

	u_1	u_2	u_3	u_4
u_1	0	1	0	0
u_2	$\frac{1}{3}$	0	$\frac{1}{3}$	$\frac{1}{3}$
u_3	0	$\frac{1}{2}$	0	$\frac{1}{2}$
u_4	0	$\frac{1}{2}$	$\frac{1}{2}$	0

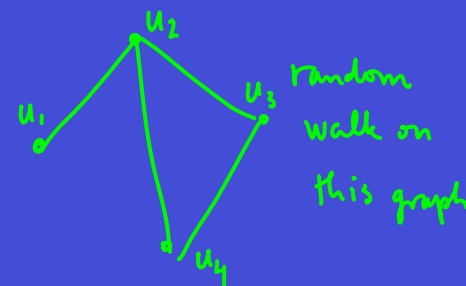
$$P = \begin{pmatrix} 0 & 1 & 0 & 0 \\ \frac{1}{3} & 0 & \frac{1}{3} & \frac{1}{3} \\ 0 & \frac{1}{2} & 0 & \frac{1}{2} \\ 0 & \frac{1}{2} & \frac{1}{2} & 0 \end{pmatrix}$$

starting in distribution σ on Ω (given by a vector)
the next distrib. is $\sigma \cdot P$

This course: state space is finite

Definition of Markov chains

- State space Ω
- Transition matrix P , of size $|\Omega| \times |\Omega|$
 - $P(x,y)$ is the probability of getting from state x to state y
 - $\sum_{y \in \Omega} P(x,y) = 1$ a prob. distribution
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$$\Omega = \{u_1, u_2, u_3, u_4\}$$

	u_1	u_2	u_3	u_4
u_1	0	1	0	0
u_2	$\frac{1}{3}$	0	$\frac{1}{3}$	$\frac{1}{3}$
u_3	0	$\frac{1}{2}$	0	$\frac{1}{2}$
u_4	0	$\frac{1}{2}$	$\frac{1}{2}$	0

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

starting in distribution σ on Ω (given by a vector)
the next distrib. is $\sigma \cdot P$

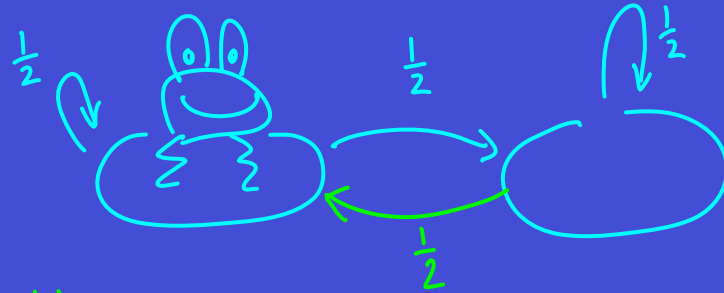
→

1st step: $\sigma \cdot P = \sigma \cdot P$
2nd step: $(\sigma \cdot P) \cdot P = \sigma \cdot P^2$
3rd step: $((\sigma \cdot P) \cdot P) \cdot P = \sigma \cdot P^3$
t-th step: $\sigma \cdot P^t$

This course: state space is finite

Definition of Markov chains

Example: frog on lily pads



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

$$\sigma = (1, 0)$$

starting on the 1st lily pad

$$\text{2nd } \sigma = \left(\frac{1}{2}, \frac{1}{2}\right)$$

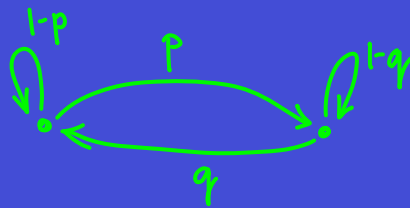
Definition of Markov chains

Stationary distribution $\frac{1}{4}$ (on states in Ω):

$$\pi \cdot P = \pi$$
$$\sum_{x \in \Omega} \pi_x = 1 \quad (\text{a distribution})$$

Example: frog with unfair coins

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



we are looking for $\pi = (\pi_1, \pi_2)$ s.t. $\pi \cdot P = \pi$

$$p, q \in (0, 1)$$

$$\pi_1 = \frac{q}{p+q}$$

$$\pi_2 = \frac{p}{p+q}$$

normalizing constant Z
(the partition function)

will often write

$$\pi_1 \propto q$$

$$\pi_2 \propto p$$

proportional to

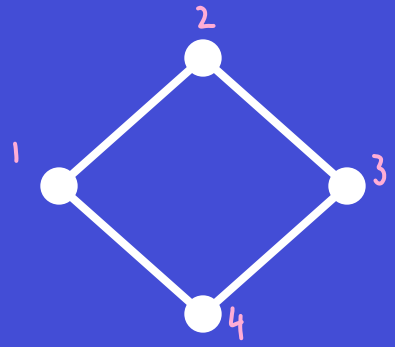
$$(\pi_1, \pi_2) \cdot P = \left(\underbrace{(1-p)\pi_1 + q\pi_2}_{= \pi_1}, \underbrace{p\pi_1 + (1-q)\pi_2}_{= \pi_2} \right)$$

Definition of Markov chains

Other examples:

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

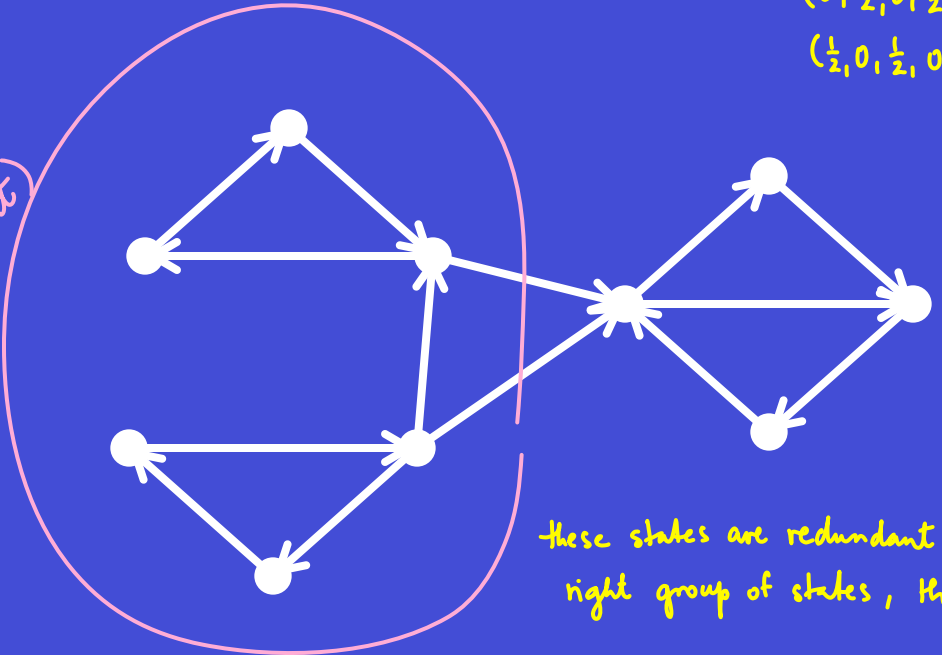
trouble: periodic (will not converge to stationary)



is $(\frac{1}{4}, \frac{1}{4}, \frac{1}{4}, \frac{1}{4})$ stationary? YES

if start in $(1, 0, 0, 0)$, then
 $(0, \frac{1}{2}, 0, \frac{1}{2})$
 $(\frac{1}{2}, 0, \frac{1}{2}, 0)$

trouble: reducible (some states are redundant)



st. distr. is $(0, 0, 0, 0, 0, 0, ? ? ? ?)$
 left states

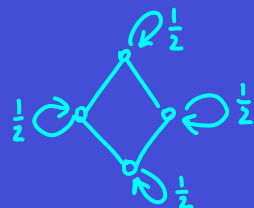
these states are redundant; once we get to the right group of states, then never come back

Definition of Markov chains

- aperiodic

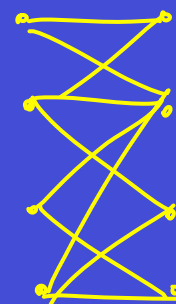
$$\forall x, y \in \Omega : \gcd \{ t \mid P^t(x, y) > 0 \} = 1$$

quick fix:



all original prob. scaled by $\frac{1}{2}$

LAZY MC (LAZY variant of the original MC)



bipartite
→ periodic

- irreducible

there exists a sequence of transitions from every state x to every state y

$$\forall x, y \in \Omega : \exists t \geq 0 : P^t(x, y) > 0$$

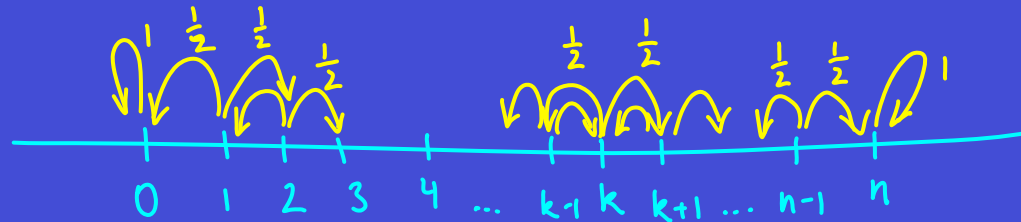
- ergodic: a MC that is both aperiodic and irreducible

Thm (3.6 in Jerrum): An ergodic MC has a unique stationary distribution. Moreover, it converges to it as time goes to ∞ .

Gambler's Ruin

A gambler has k dollars initially and bets on (fair) coin flips.
If she guesses the coin flip: gets 1 dollar.
If not, loses 1 dollar.
Stops either when 0 or n dollars.

Is it a Markov chain ?



How many bets until she stops playing ?

What are the probabilities of 0 vs n dollars ?

Gambler's Ruin

Stationary distribution ?

stat. distr: $\frac{1}{2}$ 0 0 0

\circ	\circ	\circ	\circ
0	1	2	3

not unique!

... 0 $\frac{1}{2}$
 \circ
2

$(p, 0, 0, 0, \dots, 0, 1-p)$ is stationary for any $p \in [0, 1]$

Gambler's Ruin

Starting with k dollars, what is the probability of reaching n ?

p_l - the probability of reaching n (not 0) when starting with l dollars

$\frac{k}{n}$

$$p_n = 1$$

$$p_0 = 0$$

$$p_l = \frac{1}{2} \cdot p_{l+1} + \frac{1}{2} \cdot p_{l-1}$$

for $1 \leq l \leq n-1$

↑
guessed
right

↑
not
right

solve the system of lin. equations

$$p_l = \frac{l}{n} \quad \forall l \in \{0, \dots, n\}$$

verify it:

$$p_0 = \frac{0}{n} = 0 \quad \checkmark$$

$$p_n = \frac{n}{n} = 1 \quad \checkmark$$

$$\begin{aligned} p_l &= \frac{l}{n} = \frac{1}{2} \cdot \frac{l+1}{n} + \frac{1}{2} \cdot \frac{l-1}{n} \\ &= \frac{1}{2} \cdot p_{l+1} + \frac{1}{2} \cdot p_{l-1} \end{aligned}$$

What is the probability of reaching 0?

$\frac{n-k}{n}$

Gambler's Ruin

How many bets until 0 or n?

f_k - expected # steps to get to 0 or n, starting from k

$$f_0 = 0$$

$$f_n = 0$$

$$f_l = 1 + \frac{1}{2} \cdot f_{l+1} + \frac{1}{2} \cdot f_{l-1} \quad \text{for } l \in \{1, 2, \dots, n-1\}$$

$$f_l = l(n-l)$$

Coupon Collecting

- n coupons $1, 2, \dots, n$
- each time get a coupon at random, chosen from $1, \dots, n$ (i.e., likely will get some coupons multiple times)
- how long to get all n coupons?

notice: stat. distr.:

$(0, 0, 0, \dots, 0, 1)$

Is this a Markov chain?

Option 1:

$$\Omega = \mathcal{P}(\{1, 2, \dots, n\}) \quad \text{power set}$$

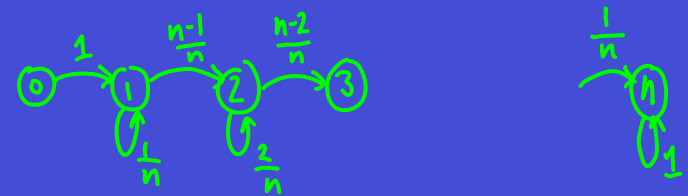
$$P(S_1, S_2) = \begin{cases} \frac{1}{n} & \text{if } S_2 = S_1 \cup \{a\}, a \notin S_1 \\ \frac{|S_1|}{n} & \text{if } S_1 = S_2 \\ 0 & \text{o/w} \end{cases}$$

Option 2:

$$\Omega = \{0, 1, \dots, n\}$$

(how many coupons have so far)

$$P(i, j) = \begin{cases} \frac{n-i}{n} & \text{if } j = i+1 \\ \frac{i}{n} & \text{if } i = j \quad \text{self-loop} \\ 0 & \text{o/w} \end{cases}$$



Coupon Collecting

How long to get all n coupons?

τ = ^{random variable for} # steps until collect all n coupons

WANT: $E[\tau]$

τ_k = ^{random variable for} # steps (of the MC) until collect k distinct coupons

$\hat{=}$
 $E[\tau_n]$

$$\tau_n = \tau_1 - \tau_0 + \tau_2 - \tau_1 + \tau_3 - \tau_2 + \dots + \tau_n - \tau_{n-1}$$

notice $\tau_0 = 0$

(no steps to get 0 coupons)

$\tau_{k+1} - \tau_k$ = ^{random var.} # steps needed after having k coupons until collecting $k+1$ coupons

$$E[\tau_n] = E[\tau_1 - \tau_0] + E[\tau_2 - \tau_1] + \dots + E[\tau_n - \tau_{n-1}] \quad (\text{linearity of expectation})$$

$$E[\tau_{k+1} - \tau_k] = ?$$

$$\frac{1}{\frac{n-k}{n}} = \frac{n}{n-k}$$

(intuitively, expect $\frac{1}{p}$ attempts to make to see an event w. prob. p to happen)

$$= \frac{n}{n-k} \quad \text{see next slide}$$

$$E[\tau_n] = \frac{n}{n} + \frac{n}{n-1} + \frac{n}{n-2} + \dots + \frac{n}{3} + \frac{n}{2} + \frac{n}{1} = n \sum_{e=1}^n \frac{1}{e} \sim n \cdot \underbrace{\log n}_{\text{natural log (ln)}}$$

Coupon Collecting

How long to get all n coupons?

$$\text{let } p = \frac{n-k}{n}$$

the prob. of getting the $(k+1)$ -st coupon, having already collected k ,
after trying l times and not succeeding $l-1$ times:

$$(1-p)^{l-1} \cdot p = \frac{1-p}{p} \cdot p = 1-p$$

where $q = 1-p$

$$E[\tau_{k+1} - \tau_k] = \sum_{l=1}^{\infty} (1-p)^{l-1} \cdot p \cdot l = p \cdot \sum_{l=0}^{\infty} q^l \cdot (l+1)$$

$$\begin{aligned} \sum_{l=0}^{\infty} q^l \cdot (l+1) &= q^0 + q^1 + q^2 + q^3 + \dots \\ &\quad + q^1 + q^2 + q^3 + \dots \\ &\quad \quad + q^2 + q^3 + \dots \\ &\quad \quad \quad + q^3 + \dots \\ &\quad \quad \quad \quad + \dots \end{aligned}$$

$$= \frac{1}{p} (1 + q + q^2 + q^3 + \dots)$$

$$= \frac{1}{p} \cdot \frac{1}{1-q} = \frac{1}{p^2}$$

$$\begin{aligned} &= \frac{1}{1-q} = \frac{1}{p} \\ &= q \cdot \frac{1}{1-q} = \frac{q}{1-q} \\ &= q^2 \cdot \frac{1}{1-q} = \frac{q^2}{1-q} \\ &= q^3 \cdot \frac{1}{1-q} = \frac{q^3}{1-q} \end{aligned}$$

Coupon Collecting

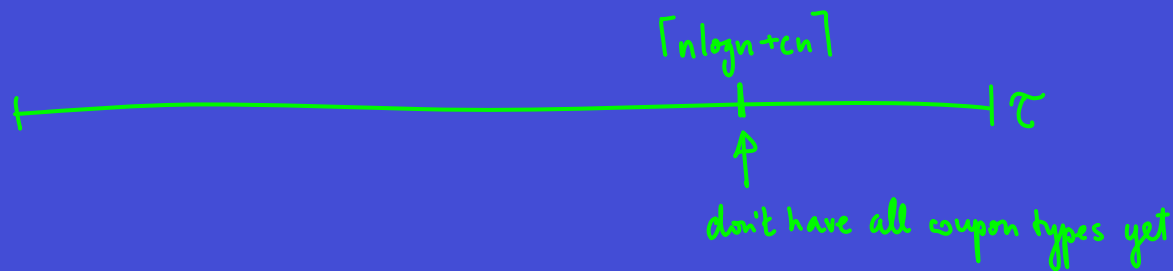
Proposition:

Let τ be the random variable for the time needed to collect all n coupons. Then,

$$\Pr(\tau > d n \log n + c n) \leq e^{-c}.$$

Let A_i be the event that we have not collected the i -th coupon type within the first $\lceil n \log n + cn \rceil$ steps.

$$\Pr(\tau > \lceil n \log n + cn \rceil) = \Pr(\exists i \text{ s.t. } A_i \text{ happened}) \leq \sum_{i=1}^n \Pr(A_i)$$



$$\begin{aligned} \Pr(A_i) &= \left(1 - \frac{1}{n}\right)^{\lceil n \log n + cn \rceil} \\ &= n \left(1 - \frac{1}{n}\right)^{\lceil n \log n + cn \rceil} \\ &\leq e^{-c} \end{aligned}$$

(recall $\log n$ natural log)

Random Walk on a Hypercube

How many expected steps until we get to a random state?
(I.e., what is the mixing time?)

Ω = all bin. strings of length n

move: choose a random position $k \in \{1, \dots, n\}$
flip k -th bit

lazy: move: w. prob. $\frac{1}{2}$ do nothing
else

stat. distr.:

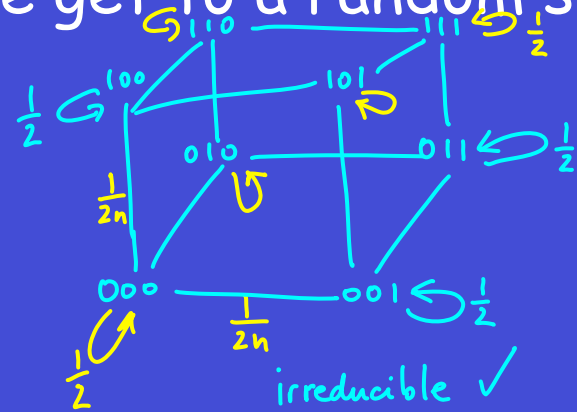
$$\pi(s) = \frac{1}{2^n}$$

verify:

$$\pi \cdot P = \pi$$

chance of getting to s_1 , assuming starting in every state w. prob. $\frac{1}{2^n}$:

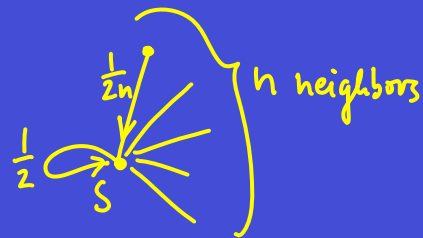
$$n \cdot \frac{1}{2^n} \cdot \frac{1}{2n} + \frac{1}{2^n} \cdot \frac{1}{2} = \frac{1}{2^n}$$



irreducible ✓
not aperiodic!

now OK

(ergodic
⇒ unique stationary
distr.)



Random Walk on a Hypercube

How many expected steps until we get to a random state?
(I.e., what is the mixing time?)

rephrased MC step:

randomly choose a position $k \in \{1, \dots, n\}$
w. prob. $\frac{1}{2}$ flip the k -th bit
else do nothing

Meaning: after selecting k , the k -th bit will be random
after selecting all positions, all bits will be random
(i.e. coupon collector!)

Hence: need $\sim n \log n$ steps