4 May 2022 Mixing time and coupling
Recall Metroplis-Hastings... (MCMC)
Finite set $X$ (think: $\mathbb{Z}$ expeneatiolly large
e.g. $Z=\{$ functions $V(\sigma) \rightarrow[q]\}$ \}

Distribution $\pi(x)=\frac{w(x)}{\sum_{y \in \bar{X}} w(y)}$ that we wish to sample from.
Strategy: Run Markov chain with transition matrix $P$ such feat $\pi P=\pi$ until it "mixes", meaning state distrib gets close to $\pi$.
Then output the current state; it will be an approximate sample from $\pi$.
Metroplis-Hastings: given a different Markov chain $K$ on I with $K_{x y}=K_{y x}$. $\Leftrightarrow$ uniform district is stationary for $K$.) Want to medity $K$ into a Markov chain $P$ such that (a) $\pi P=\pi$
(10) Easy to simulate one state transition of $P$.

$$
P_{x y}=K_{x y} \cdot \frac{\min \{w(x), w(y)\}}{w(x)} .
$$

$P:$ Given current state $x$

1. Randan sample "proposed react state" $y$ with pars $K_{x y}$.
2. Transition to $y$ with prob $\frac{\min \{w(x) w(y)\}}{w(x)}$.

Assumption is that bath $K$ and $\rho$ will be implemented lily subroutines that input state $x$ and output state $y$ with probability $K_{x y}$ os $P_{x y}$ respectively.
Eg. Metroplis-Hastings for sampling random proper $q$-coloring of soph $G$.

$$
\begin{aligned}
& X=\{\text { functions } \quad v(G) \rightarrow[q]\} \\
& w(x)=\left\{\begin{array}{llll}
1 & \text { if } & x(u) \neq x(v) & \forall \text { edge }(u, v) \\
\phi & \text { if } & x(u)=x(v) & \text { for some edge }(u, v)
\end{array}\right.
\end{aligned}
$$

For $K$, we will use a very simple subroutine. In state $x$, pick a unit tandem vertex $v \in V(\sigma)$ a unit roudom color $c \in[q]$ and modify the cole of $v$ to equal $c$.
ie. output state $y: V(G) \rightarrow[q]$ sit.

$$
y(u)=\left\{\begin{array}{cc}
c & \text { if } u=v \\
x(u) & \text { if } u \neq v
\end{array}\right.
$$

Def Hamming distance $d(x, y)$ is

$$
\begin{gathered}
d(x, y)=\text { \# }\{v \mid x(v) \neq y(v)\} \\
K_{x y}=\left\{\begin{array}{lll}
0 & \text { if } d(x, y)>1 \\
\frac{1}{n \cdot g} & \text { if } & d(x, y)=1 \\
\frac{1}{q} & \text { if } & d(x, y)=0
\end{array} \quad K_{x y}=K_{y x}\right. \text { blaming dist is }
\end{gathered}
$$

metropild - Hastings for this particular $K$ says.
Given current stance $x$. (Assume x a proper coloring.)

1. Sample $y$ with probability $K_{x y} . \Rightarrow w(x)=1$ Pick random $v \in V(G), \quad c \in[q]$, try setting $y=\{x$ with $\vee$ recolored as $c\}$.
2. Transition to $y$ with probability $\frac{\min \{w(x), w(y)\}}{w(x)}$,

$$
\begin{aligned}
\operatorname{Prob}(\text { trans }) & =\min \{1, w(y)\}=w(y) \\
& =\left\{\begin{array}{cl}
1 & \text { if } y \text { is proper coloring } \\
\varnothing & \text { if wot. }
\end{array}\right.
\end{aligned}
$$

Test if color $C$ matches the color of a neighbor of $u$
If so, vo transition. (Output $x$ ) If not, output $y$.

After how mary stops can we terminate the Markov chain, safe in the knowledge that its marginal distribution is near stationary?

Def= If $\pi, \pi^{\prime}$ are pros. distints, their total variation distance is

$$
\begin{aligned}
\left\|\pi-\pi^{\prime}\right\|_{T V} & =\max _{S \subseteq \bar{X}}\left\{\pi(S)-\frac{1}{\pi(S)}\right\} \\
& =\frac{1}{2}\left\|\pi-\pi^{\prime}\right\|_{1}
\end{aligned}
$$

For us, "near stationary" will mean $\varepsilon$-close to the stationary distend in total variation distance.

Def. A coupling of two Markov chains gm, M2 is a probability distil. over sequences of pairs $\left(X_{t}, X_{t}^{\prime}\right)$ such that the marghal distrib of $X_{0}, X_{1}, X_{2}, \ldots$ is a random sample from $M_{1}$,

$$
x_{0}^{\prime}, x_{1}^{\prime}, x_{2}^{\prime} \cdots \cdots \cdots M_{2}
$$

Lemma. If $\left\{\left(X_{t}, x_{t}^{\prime}\right)\right\}$ is a coupling of two Markov chains with same state set $X$, same
transition matrix $p$ and with $X_{0}^{\prime}$
drown from $\pi$, stationary distils of $P$ then the marginal distrib of $X_{t}$, called $\pi_{t^{\prime}}$ Satisfies

$$
\left\|\pi_{t}-\pi\right\|_{\pi v} \leq \quad \operatorname{Pr}\left(X_{t} \neq X_{t}^{\prime}\right) .
$$

Proof. Since $X_{0}^{1}$ is down from $\pi$ and $x_{0}^{\prime}, x_{1}^{\prime}, \ldots . f_{0}$ follows transition matrix $P$ whose stationary dirtrib is $\pi$, it means $X_{t}^{\prime} \sim \pi$ for all $t>0$. (Induct on $t_{1}$ )
By definition of TV distance, to prove inequality in Lemma we must show

$$
\begin{aligned}
& \pi_{t}(s)-\pi(s) \leqslant \operatorname{Pr}\left(x_{t} \neq x_{t}^{\prime}\right) \quad V s \subseteq \bar{X} \\
& \operatorname{Pr}\left(X_{t} \in S\right)-\operatorname{Pr}\left(X_{t}^{\prime} \in S\right)
\end{aligned}
$$

If $X_{t} \in S$ then either $X_{t}^{\prime} \in S$ or $X_{t} \neq X_{t}^{\prime}$. $\Downarrow$ (union hod)

$$
\begin{gathered}
\operatorname{lr}\left(x_{t} \in S\right) \leqslant \operatorname{Pr}\left(x_{t}^{\prime \prime} \in S\right)+\operatorname{Pr}\left(x_{t} \neq x_{t}^{\prime}\right) \\
\operatorname{Pr}\left(x_{t} \in S\right)-\operatorname{Pr}\left(x_{t}^{\prime} \leqslant 5\right) \leqslant \operatorname{Pr}\left(x_{t} \neq x_{t}^{\prime}\right)
\end{gathered}
$$

