

Lecture 12

①

Flow Encodings:

- Recall from last time: If we route commodity $x \rightarrow y$ through unique path γ_{xy} , and stationary distn' is uniform, then for the flow to result in polynomial mixing time we need:

$$|\text{paths}(e)| \leq \text{poly}(n) \cdot |\Omega|, \forall e \quad (*)$$

\nwarrow \swarrow
 $\# \text{ paths through edge } e$ $\text{natural size of problem}$

Remarks: $(*)$ is ^{still} required in the average sense, if we use several paths for each commodity

\hookrightarrow if stationary distn' not uniform, $(*)$ can be suitably reweighted

- Immediate Execution Issue: $|\Omega|$ is unknown, and is often what we are after.

\rightarrow so need to check $(*)$ implicitly.

\rightarrow Idea: charge every element of $\text{paths}(e)$ to some point in Ω
 \hookrightarrow each point in Ω is charged once!

Def: [flow Encoding]: An encoding for a flow f (that only uses single path γ_{xy} for each commodity $x \rightarrow y$) is a set of functions $\{\eta_e: \text{paths}(e) \rightarrow \Omega\}_e$ such that:

1. η_e is injective, for all e
2. $\pi(x) \cdot \pi(y) \leq b \cdot \pi(z) \cdot \pi(\eta_e(x, y))$, $\forall (x, y), \forall e \in \gamma_{xy}, e = (z, z')$

- automatically true with $b=1$, if stationary distn' is uniform

- in general, it says that the η_e 's are weight-preserving up to a factor b (want this as small as possible)

Remark: Sometimes it is ok if the η_e 's are not perfect injections, but we use a little "extra information" to invert them (see proof of next claim).

Claim: If there is an encoding for f as above, then $p(f) \leq b \cdot \max_{P(z,z') > 0} \frac{1}{P(z,z')}$.

Proof: For arbitrary $e = (z, z')$:

$$f(e) = \sum_{(x,y) \in \text{paths}(e)} \pi(x) \pi(y) \leq b \sum_{(x,y) \in \text{paths}(e)} \pi(z) \cdot \pi(n_e(x,y)) \leq b \cdot \pi(z) = b \cdot \frac{C(e)}{P(z,z')}$$

$$\Rightarrow \frac{f(e)}{C(e)} \leq \frac{b}{P(z,z')}$$

□

↑ we used the fact that n_e is an injection.
(if n_e becomes an injection w/ a little "extra information" this results in an additional factor in this inequality)

• Example: Lazy RW on $\{0,1\}^n$

Last time: - analyzed RW by splitting flow evenly on all shortest paths between x, y for all (x, y) ; then appealed to symmetries of the cube to analyze the flow, and we used that $|S| = 2^n$.

This time: Pretend we don't know that $N = |S| = 2^n$, and use flow encodings.

• γ_{xy} : left-to-right bit fixing path (e.g. $010 \xrightarrow{L} 110 \xrightarrow{L} 100 \xrightarrow{L} 101$)

• clearly, $l(f) = n$

• analyze $p(f)$ using flow encodings:

→ suppose $e = (z, z')$, where z, z' are different at i

→ let $\gamma_{xy} \ni e$ for some pair x, y

• clearly x agrees with z in bits $i \dots n$ and y agrees with z in bits $1 \dots i-1$

• so can define $n_e(x, y) = x_1 x_2 \dots x_{i-1} y_i y_{i+1} \dots y_n$

• n_e is an injection because given $n_e(x, y)$ and z I can invert x, y .

• also, stationary distn' is uniform, so encoding is weight-preserving with $b=1$

So n_e is a flow encoding, so claim gives:

$$\rho(f) \leq \max_{P(z,z') > 0} \frac{1}{P(z,z')} = 2n$$

so obtained $\rho(f) \cdot \ell(f) \leq 2n^2$ (up to a constant factor this is the same we got last time; but we didn't use the

cube's symmetries or size of $|S|$)

Example 2: Matchings in Unweighted Undirected Graphs

[Jerrum + Sinclair '89: Approximating the Permanent, SICOMP 18, 1989]

Input:

- $G = (V, E)$: unweighted, undirected
- parameter $\lambda \geq 1$ (essentially same technology is used for $\lambda < 1$)
- $\Omega = \{\text{matchings of } G\}$

Goal: Sample from Gibbs distn' on Ω :

$$\pi(M) = \frac{1}{Z} \lambda^{|M|} \quad \leftarrow \text{\#edges in } M$$

$$Z = Z(\lambda) = \sum_k m_k \lambda^k, \quad m_k = \text{\#matchings with } k \text{ edges.}$$

\hookrightarrow matching polynomial.

$\#P$ -complete for all $\lambda > 0$

- Statistical Physics Motivation: monomer-dimer model

(edges in matching: diatomic molecules)
(vertices: monatomic molecules)

Markov Chain: \rightarrow make it lazy

- \rightarrow use 3 kinds of transitions: edge addition, edge deletion, edge exchange
- \rightarrow use Metropolis Rule to make sure that stationary distn' matches Gibbs distribution.

MC step if at state M :

- with prob. $1/2$, stay at M (laziness)
- o.w. choose an edge $e = (u, v) \in E$ u.a.r.
- if u, v unmatched in M , go to $M+e$ (edge addition)
- if $e \in M$, go to $M-e$, w/ probability $1/2$ (edge deletion)
and stay at M w/ prob $1-1/2$
- if exactly one of u or v is matched in M , go to $M+e-e'$ (edge exchange)
where e' is edge adjacent to u or v
- if both u, v matched, do nothing

Chain follows metropolis rule, hence stationary distn' is gibbs distn'.

Flow:

• Let x, y be matchings; color x red, y blue.

• want to specify γ_{xy}

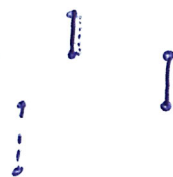
• superimpose x, y : $x+y$

↳ comprises of: simple paths of alternating red, blue edges

- simple cycles of $-//-$

- edges that are both red & blue

e.g.



solid edge: red
dashed edge: blue

master
index

fix (for our analysis) a total ordering of all simple paths & cycles that are subgraphs of G and designate a "start vertex" in all of them (the start vertex of a path should be an endpoint)

↳ this ordering induces an ordering on the paths and cycles that appear on $x+y$

• define δ_{xy} as follows:

- process paths & cycles in ~~any~~ order specified by master index:
- to process a path: let e_1, e_2, \dots, e_f be its edges where e_1 is adjacent to start vertex; process it via the following transitions:
 - i) if e_1 is red, remove it, then exchange e_3 for e_2 , exchange e_5 for e_4 etc; if e_f is blue, add e_f in the end.
 - ii) if e_1 is blue, exchange e_2 for e_1 , e_4 for e_3 , etc; if e_f is blue, add it in the end.
- to process a cycle: let e_1, e_2, \dots, e_f be its edges, where e_1 is red edge adjacent to start vertex; process cycle as follows: remove e_1 , exchange e_3 for e_2 , e_5 for e_4 , etc; finally add e_f

• flow encoding: - suppose ^{transition} $t = (z, z')$ corresponds to edge exchange; we proceed to define n_t (for other types of transition the encoding is similar)

- suppose t involves exchanging e' for e , where $e, e' \in E$
- let $x, y \in \Omega$ be matchings for which the path δ_{xy} uses transition $t = (z, z')$
- define

$$n_t(x, y) = \begin{cases} x \oplus y \oplus (zuz') & \text{if } \delta_{xy} \text{ uses } t \text{ when processing a path (case 1)} \\ x \oplus y \oplus (zuz') \setminus \{e_1\} & \text{if } \delta_{xy} \text{ uses } t \text{ when processing a cycle whose first (red) edge is } e_1 \text{ (case 2)} \end{cases}$$

- Demystifying $\eta_t(x, y)$:

- Suppose transition $t = (z, z')$ is used in δ_{xy} when processing a path or cycle S in $x+y$

- easy to check: z, z' differ only in e, e' (z has e & not e' , z' has e' & not e)

◦ z, z' :- all edges that are colored both red and blue in $x+y$ appear in z, z'

- for all connected components of $x+y$ that precede S in the total ordering z, z' have only kept the blue edges of these components

- for all connected components of $x+y$ following S in the total ordering z, z' have the red edges of these components

- with regards to S :

◦ z has kept all blue edges before e , except all red edges after (including) e

◦ z' has kept all blue edges before (including) e' and all red edges after e' , except e

- So easy to check that $\eta_t(x, y)$ is a matching (removing e in the case that S is a cycle is crucial for this to hold)

- Now is it an injection?

- If we knew whether " $\setminus \{e, e'\}$ " exists in definition of $\eta_t(x, y)$,

we can recover $x \oplus y$ as follows:

$$x \oplus y = \eta_t(x, y) \oplus (z \cup z'), \text{ for case 1}$$

$$x \oplus y = (\eta_t(x, y) \cup \{e, e'\}) \oplus (z \cup z'), \text{ for case 2}$$

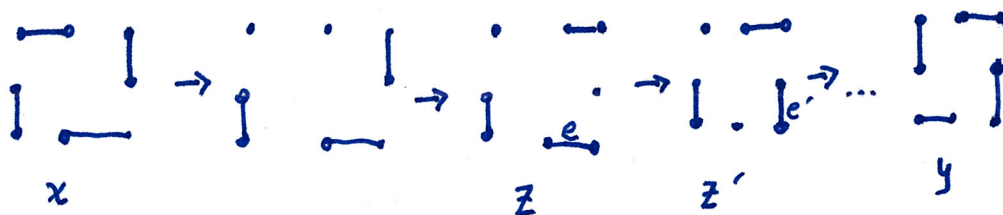
- and if we have $x \oplus y$ we can obtain x and y by looking at $x \oplus y, z, z'$

- edges in $z \setminus x \oplus y$ are in both x & y
- can identify the order in which the components of $x \oplus y$ were processed by looking at master ordering and $x \oplus y$ (note: the c.c. that need processing in $x \oplus y$ are the c.c.s of $x \oplus y$)
- by looking at z, z' and the connected components of $x \oplus y$ can identify which c.c. was being processed during the transition $t = (z, z')$, which c.c. were processed before, and which after
- then $z, z', x \oplus y$ will identify what edges are blue and what are red in these components (see discussion in page 6).

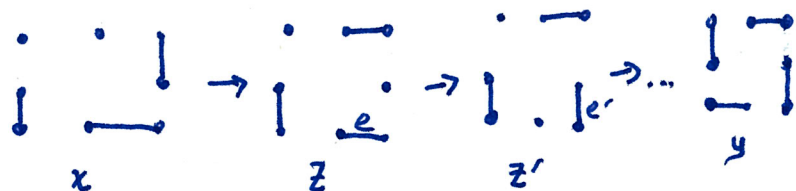
- However, cannot quite get $x \oplus y$ by looking at $\eta_t(x, y), z, z'$

e.g.

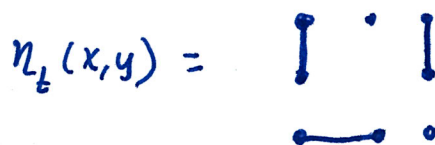
scenario 1:



vs scenario 2:



in both cases:

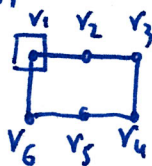


and z, z' are obviously the same.

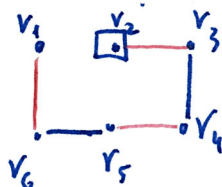
- Remedy: Problem arises because cycle and path that was missing an edge of the cycle, were processed in the same order (that was adjacent to start vertex)

→ solution: master index specifies different start vertex for the same path but w/ opposite red/blue edge alternations, if the endpoints of the path are adjacent and the corresponding cycle's start vertex is adjacent to missing edge.

e.g. suppose v_2 is the start vertex of the cycle $v_1, v_2, v_3, v_4, v_5, v_6$

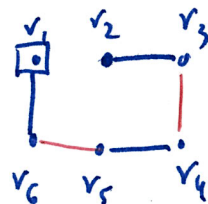


Consider the paths:



↑ choose v_2 as start vertex

and



↑ choose v_1 as start vertex

- Claim: If we use the above convention for processing paths & cycles, given $n_t(x, y)$, z and z' we can always recover $x \oplus y$ and therefore x and y .

Proof: Can figure out $x \oplus y$ except maybe one edge ^{will be missing} if the current c.c. being processed during transition $t = (z, z')$ is a cycle; but looking at z, z' we can figure out the direction in which the component was being processed and using the convention above we can tell if it was a path or a cycle.

weight-preservation?

- compare $|x| + |y|$ to $|z| + |n_z(x, y)|$
- not hard to see, that the latter has at most 2 fewer edges (possible edge deletion at beginning of cycle, 1 edge missing from current transition).

$$\Rightarrow \pi(x) \cdot \pi(y) \leq \lambda^2 \pi(z) \cdot \pi(n_z(x, y)) \quad (*)$$

- analysis of τ_{mix} : for any transition $t = (z, z')$, $P(z, z') \geq \frac{1}{2\lambda|E|}$ (**)

$$\stackrel{(*)}{\Rightarrow} \stackrel{(**)}{p(f)} \leq \lambda^2 \cdot \left(\frac{1}{2\lambda|E|} \right)^{-1} = O(\lambda^3 \cdot |E|).$$

also clearly $\ell(f) \leq |V|$ (since $|x| \leq \frac{|V|}{2}$, $|y| \leq \frac{|V|}{2}$ and every vertex in $x \oplus y$ is processed once)

$$\Rightarrow \alpha \geq \frac{1}{p(f) \cdot \ell(f)} \geq \Omega\left(\frac{1}{\lambda^3 |E| \cdot |V|}\right).$$

$$\Rightarrow \tau_x(\varepsilon) \leq O\left(\lambda^3 \cdot |E| \cdot |V| \cdot (\log \pi(x)^{-1} + 2 \log(V\varepsilon))\right)$$

- starting x ?

start from a max-weight matching (can be found efficiently)

$$\pi(x) \geq \frac{1}{|\Omega|} \geq \frac{1}{2^{|E|}}$$

$$\Rightarrow \tau_x(\varepsilon) = O(\lambda^3 |E|^2 \cdot |V|).$$

exercise (1pt): Impose bound to $O(\lambda^2 \cdot |E|^2 \cdot |V|)$.