



Rapidly Mixing Chain and Perfect Sampler
for Logarithmic Separable Concave
Distributions on Simplex

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Problem

Given

each f_i is log-concave

log-concave functions vector $f = (f_1, \dots, f_n)$

Sample space

simplex

$$\Xi = \{(x_1, \dots, x_n) \in \mathbb{Z}^n \mid x_i \geq 0, x_1 + \dots + x_n = K\}$$

Probability function

logarithmic separable
concave function

$$\pi(\mathbf{x}) = \frac{1}{C} \prod_{i=1}^n f_i(x_i)$$

where $C \stackrel{\text{def.}}{=} \sum_{\mathbf{x} \in \Xi} \prod_{i=1}^n f_i(x_i)$

log-concave function f_i

$$f_i : \mathbb{Z} \rightarrow \mathbb{R}_{++} \quad (i \in \{1, \dots, n\}),$$

$$2 \ln f_i(z+1) \geq \ln f_i(z) + \ln f_i(z+2).$$

- ex. of log-concave
- exponential dist.,
 - normal dist.,
 - positive concave.

logarithmic separable concave function

product of single variable log-concave functions.

Main results

For any **logarithmic separable concave** discrete distribution on simplex, we give two hit-and-run chains.

- One provides an **approximate sampler**.

- The **mixing time** $\tau(\varepsilon)$ satisfies

$$\tau(\varepsilon) \leq \frac{n(n-1)}{2} \ln(K\varepsilon^{-1}).$$

- Proof by **path coupling**.

- The other provides a **perfect sampler**.

- The chain is **monotone**.

- The perfect sampler is based on monotone **CFTP** (**C**oupling **F**rom **T**he **P**ast)

⇒ **Exactly** according to the stationary distribution.

Related works and applications

Related works

- A. Frieze and R. Kannan ('97)
 - log-concave on **continuous** polytope
 - ⇒ **polynomial time** mixing
- L. Lovasz and S. Vempara ('02)
 - log-concave on **continuous** polytope
 - ⇒ mixing time is $O^*(n^4)$
- D. Randall and P. Winker ('05)
 - uniform on **continuous** simplex
 - ⇒ mixing time is $\Theta(n^3 \ln n)$

Discrete is **NOT** easy

Applications

- Jackson networks
- Discretized Dirichlet Distribution

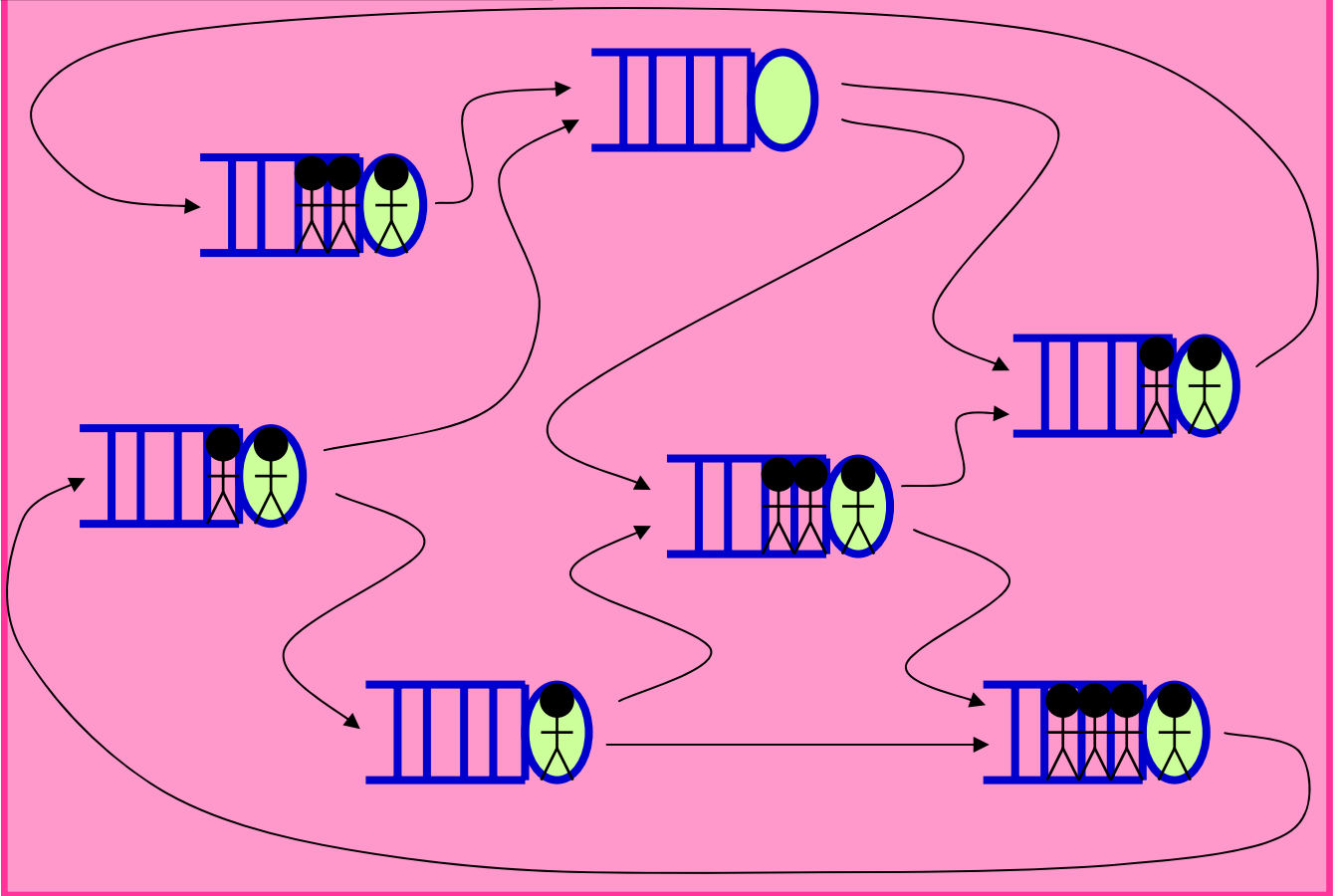
An application

- Jackson networks: Queueing network theory

⊠ : job (customer)

How long should we wait?

□□□□○ node (server)



Steady state distribution of jobs

$$\pi_J(\mathbf{x}) \stackrel{\text{def.}}{=} \frac{1}{C_J} \prod_{i=1}^n \frac{1}{\prod_{j=1}^{x_i} \min\{j, s_i\}} \left(\frac{\theta_i}{\mu_i} \right)^{x_i}$$

with given positive constant vectors

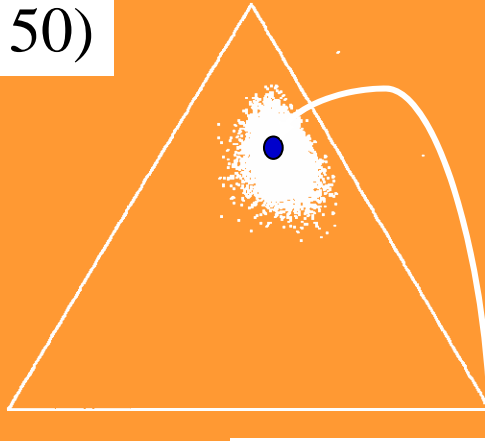
$$(\theta_1, \theta_2, \dots, \theta_n), (\mu_1, \mu_2, \dots, \mu_n) \text{ and } (s_1, s_2, \dots, s_n)$$

Another application

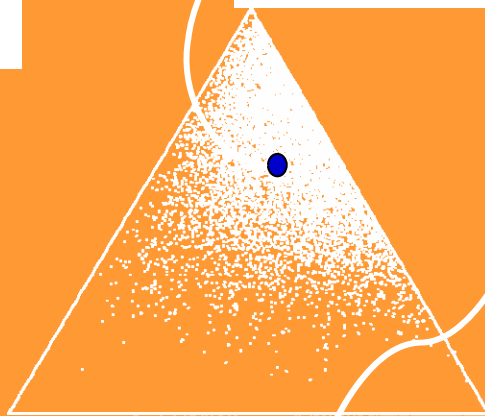
- Discretized Dirichlet Distribution

$$\pi_D \stackrel{\text{def.}}{=} \frac{1}{C_D} \prod_{i=1}^n \left(\frac{x_i}{K} \right)^{u_i-1}$$

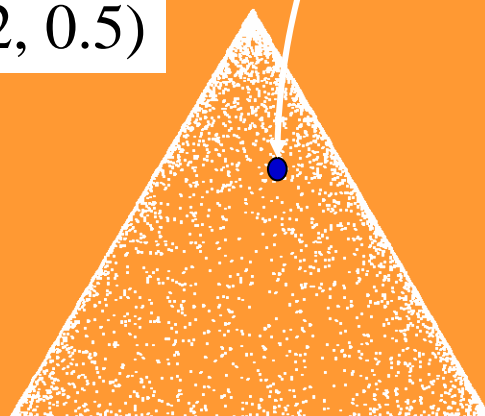
$$(u_1, u_2, u_3) = (10, 20, 50)$$



$$(u_1, u_2, u_3) = (1, 2, 5)$$



$$(u_1, u_2, u_3) = (0.1, 0.2, 0.5)$$



mean = (1/8, 2/8, 5/8)

Markov chain

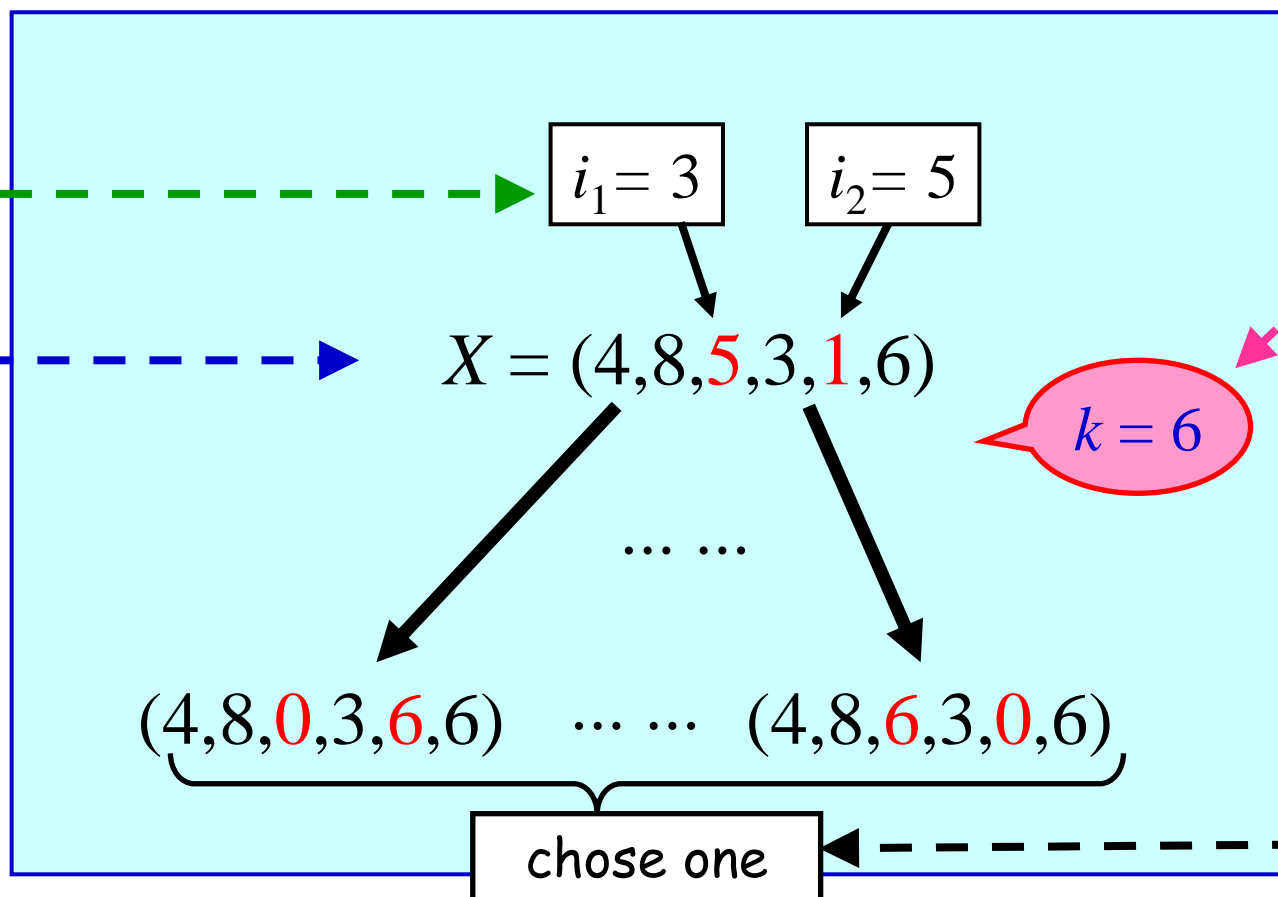
Current state: $X \in \Xi$

• Chose a pair of indices $\{i_1, i_2\}$, u. a. r.

• Set $k = X_{i_1} + X_{i_2}$

• Chose $l \in \{0, 1, \dots, k\}$ w. p. $\frac{f_{i_1}(l)f_{i_2}(k-l)}{\sum_{s=0}^k f_{i_1}(s)f_{i_2}(k-s)}$

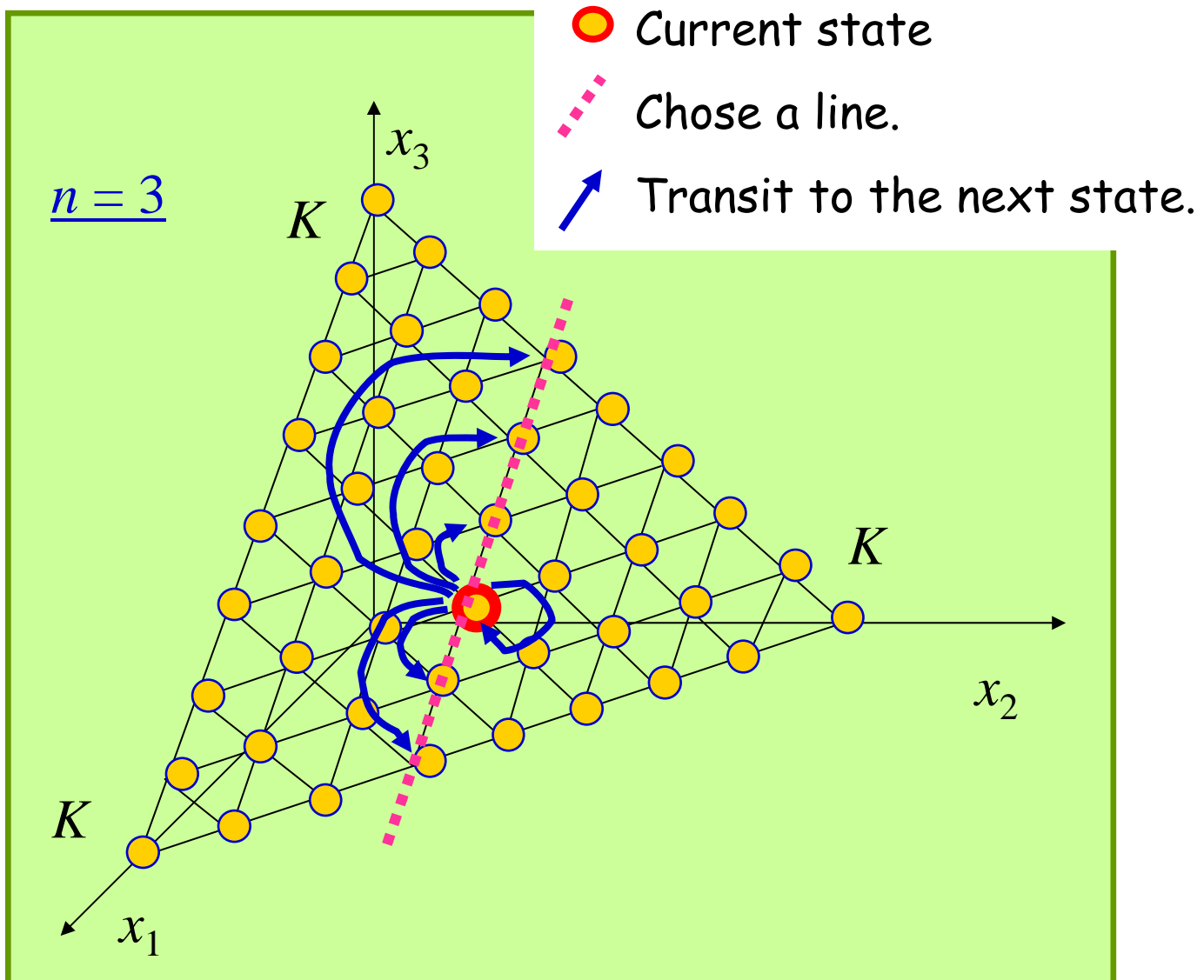
• Set $X'_i = \begin{cases} l & (i = i_1) \\ k - l & (i = i_2) \\ X_i & (\text{otherwise}) \end{cases}$



Stationary distribution

Them.

The chain is ergodic and has a unique stationary distribution which is π



Approximate sampler

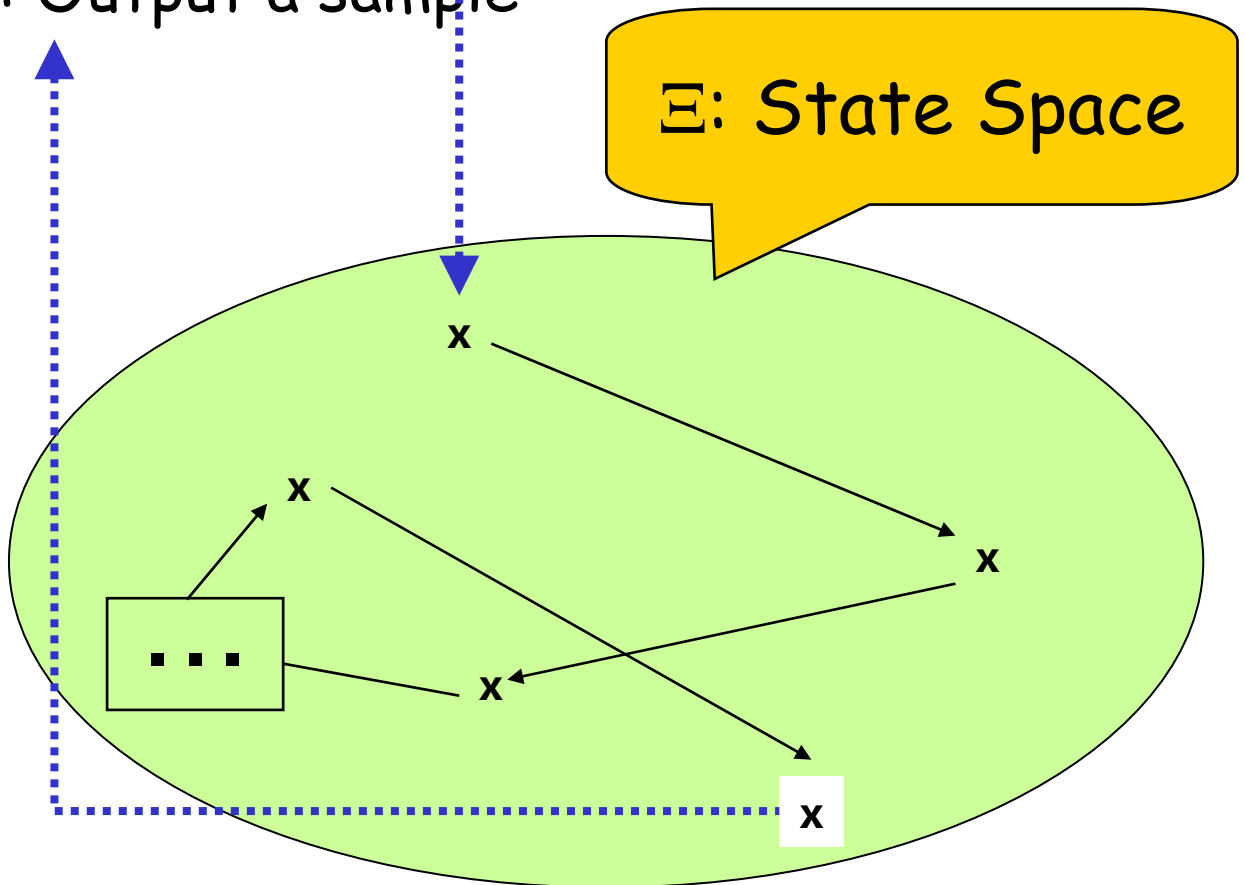
- Sampling via Markov chain

Basic idea

1: Start from an initial state

2: Make several transitions

3: Output a sample



If a chain is ergodic, it converges to a unique **stationary distribution**.

Mixing time

Them. 1

The **mixing time** $\tau(\varepsilon)$ of the chain satisfies

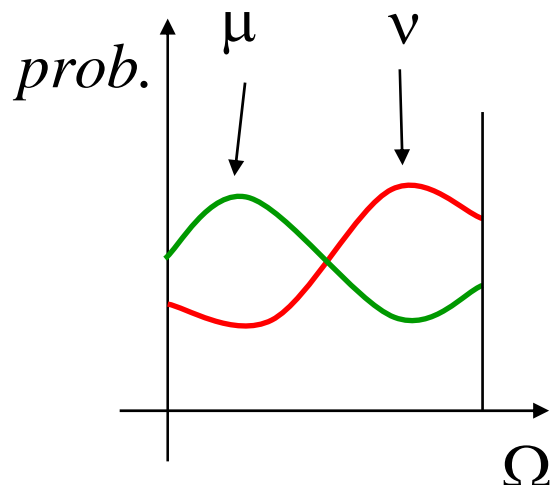
$$\tau(\varepsilon) \leq \frac{n(n-1)}{2} \ln(K\varepsilon^{-1}).$$

Total variation distance between μ and ν is defined by

$$\begin{aligned} d_{\text{TV}}(\mu, \nu) &\stackrel{\text{def.}}{=} \max_Q \left\{ \sum_{x \in Q} (\mu(x) - \nu(x)) \right\} \\ &\equiv \frac{1}{2} \sum_{x \in Q} |\mu(x) - \nu(x)| \end{aligned}$$

Mixing time of a chain is defined by

$$\tau(\varepsilon) \stackrel{\text{def.}}{=} \max_{x \in \Omega} \left\{ \min\{t \mid \forall s \geq t, d_{\text{TV}}(P_x^t, \pi) \leq \varepsilon\} \right\}$$



Proof (by path coupling)

- Graph: $G = (\Xi, E)$
- Edge: $\{X, Y\} \in E$
 $\Leftrightarrow \exists \{j_1, j_2\} (j_1 \neq j_2), X_j - Y_j = \begin{cases} 1 & (j \in \{j_1, j_2\}), \\ 0 & (\text{otherwise}) \end{cases}$
- Distance: $d_A(X, Y)$ shortest path between X and Y on G
- Joint process: $(X, Y) \mapsto (X', Y')$ cumulative dist.

Claim

$$\forall \{X, Y\} \in E, \mathbb{E}[d_A(X', Y')] \leq \left(1 - \frac{2}{n(n-1)}\right) d_A(X, Y)$$

Proof: Let $\{X, Y\} \in E$ and $\{j_1, j_2\} (j_1 \neq j_2)$ be a supporting pair. Suppose a pair $\{i_1, i_2\} (i_1 \neq i_2)$ is chosen.

Case 1. $\{i_1, i_2\} \cap \{j_1, j_2\} = \emptyset$ (neither of j_1, j_2 are chosen)

$$\triangleright d_A(X', Y') = d_A(X, Y) \equiv 1$$

Case 2. $\{i_1, i_2\} = \{j_1, j_2\}$ (both of j_1, j_2 are chosen)

$$\triangleright d_A(X', Y') = 0 \quad \leftarrow \text{w. p. } \frac{2}{n(n-1)}$$

Case 3. $|\{i_1, i_2\} \cap \{j_1, j_2\}| = 1$ (exactly one of j_1, j_2 is chosen)

$$\triangleright d_A(X', Y') = d_A(X, Y) \equiv 1$$

Them. Path coupling [Bubley and Dyer '97]

$$0 \leq \exists \beta < 1, \forall \{X, Y\} \in E, \mathbb{E}[d(X', Y')] \leq \beta \cdot d(X, Y)$$

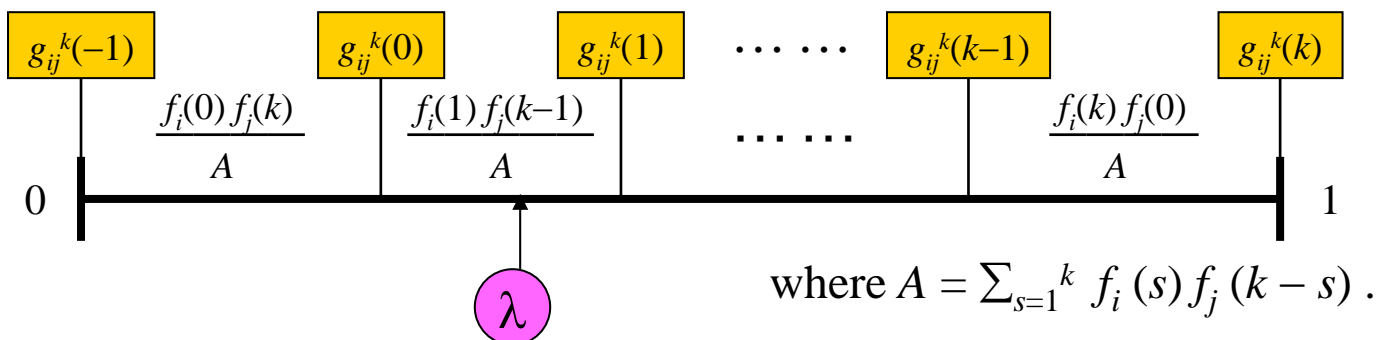
$$\Rightarrow \tau(\varepsilon) \leq (1 - \beta)^{-1} \ln(D\varepsilon^{-1}).$$

Alternating inequalities

Joint process

Simulate the chain with random number $\lambda \in [0,1)$ and **cumulative distribution function**

$$g_{ij}^k(l) \stackrel{\text{def.}}{=} \frac{\sum_{s=1}^l f_i(s) f_j(k-s)}{\sum_{s=1}^k f_i(s) f_j(k-s)}$$

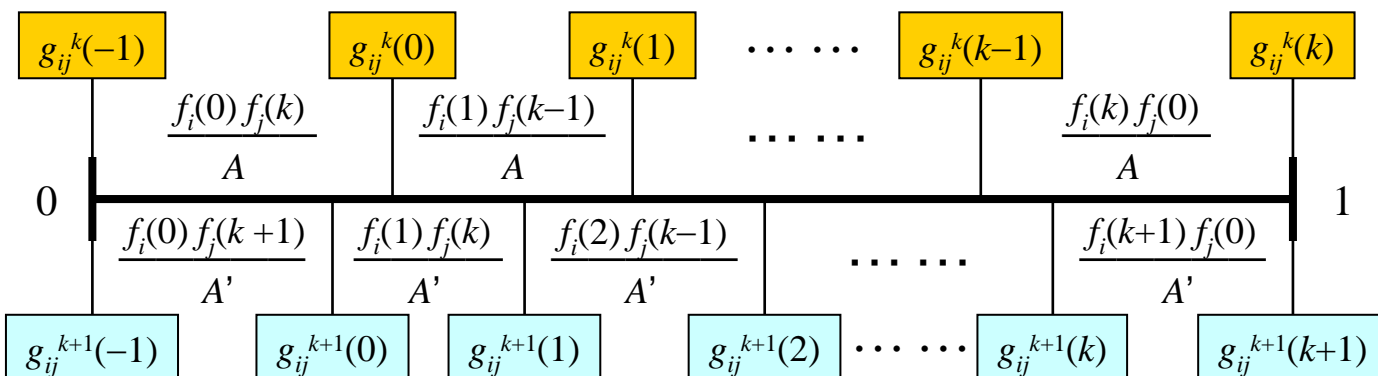


Lemma

A logarithmic separable concave function satisfies **alternating inequalities**

$$g_{ij}^{k+1}(l) \leq g_{ij}^k(l) \leq g_{ij}^{k+1}(l+1) \quad \forall l \in \{0, 1, \dots, k\}$$

for any $k \in \{0, \dots, K\}$ and for any pair $\{i, j\}$.



Another chain - for perfect sampling

modification

consecutive

- Chose a pair of indices $\{i, i + 1\}$, u. a. r.

Them. 2

The chain is **monotone**.

Equipments

- **Cumulative sum vector** c_x for $x \in \Xi$

$$c_x(i) \stackrel{\text{def.}}{=} \begin{cases} 0 & (i = 0) \\ \sum_{j=1}^i x_j & (i \in \{1, 2, \dots, n\}) \end{cases}$$

- Partial order on Ξ

$$x \succeq y \quad (x, y \in \Xi) \stackrel{\text{def.}}{\Leftrightarrow} c_x \geq c_y \quad (\forall i \in \{0, 1, \dots, n\})$$

- Max. and min.

$$\text{Max.: } x_U = (K, 0, 0, \dots, 0, 0)$$

$$\text{Min.: } x_L = (0, 0, 0, \dots, 0, K)$$

- update function

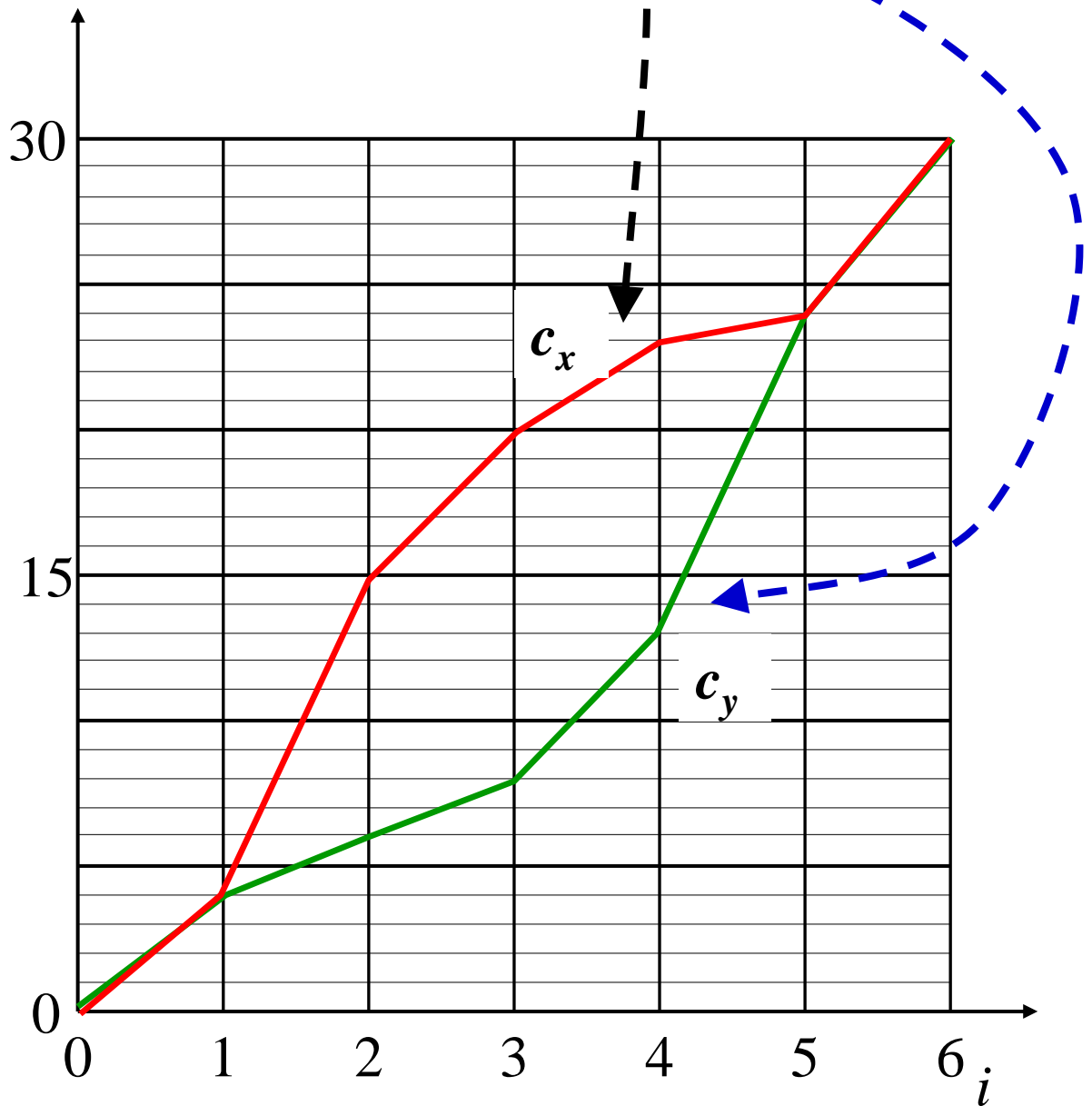
cumulative distribution function

Cumulative sum vector

Example

$$x = (4, 11, 5, 3, 1, 6)$$

$$y = (4, 2, 2, 5, 11, 6)$$



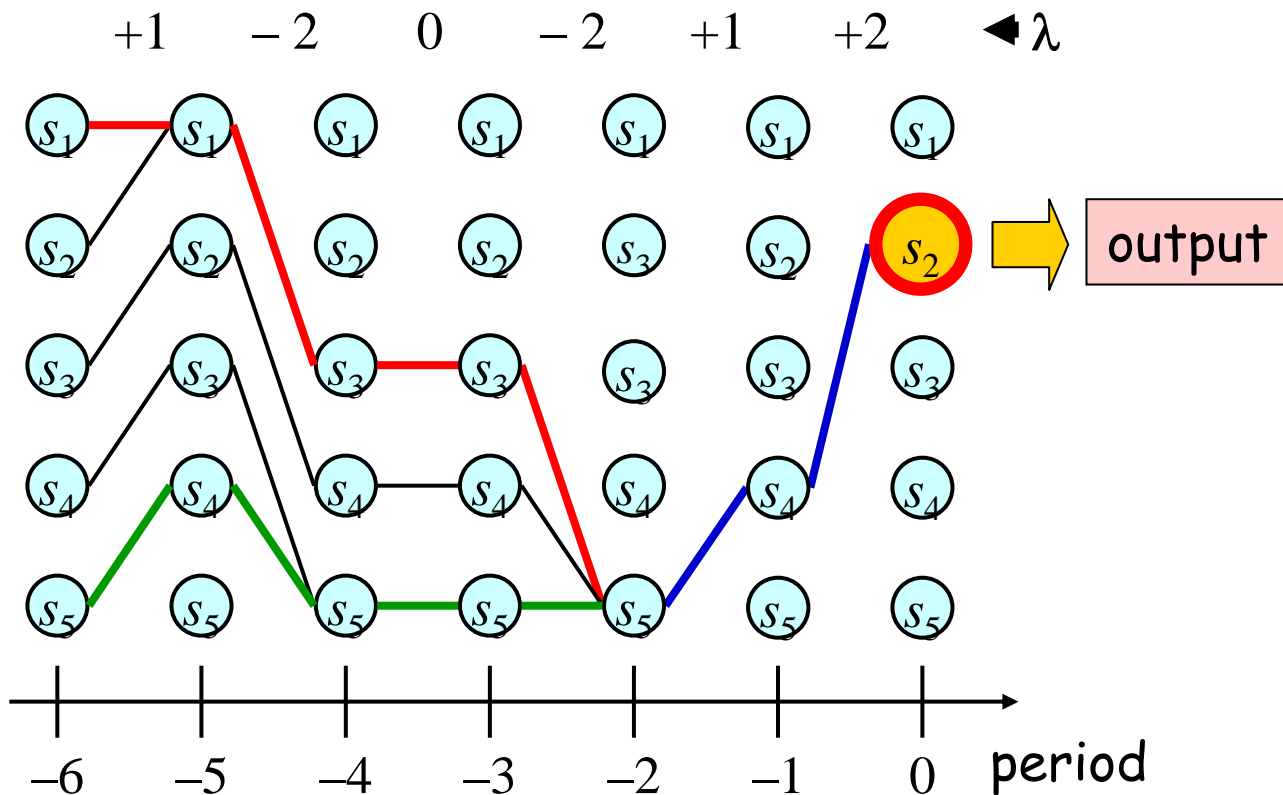
Thus $x \succcurlyeq y$

monotone CFTP [Propp and Wilson '96]

monotone Markov chain

- The state space of a chain has a **poset**.
- Any transitions preserve partial order
- The poset has a unique max and min.

monotone CFTP



Them. monotone CFTP

monotone CFTP algorithm returns a sample in probabilistic finite time, **exactly** according to the stationary distribution.

Concluding remarks

Future works

- Is the mixing time (or expected **coalescence time**) of our **monotone chain** bounded by polynomial time?
- Extension to log-concave (not separable)
 - application: universal portfolio.
- Extension to integer points on more general polytopes?
 - application: loss networks.
 - base polytope of submodular function.

Technical reports are available from

<http://www.keisu.t.u-tokyo.ac.jp/Research/techrep.0.html>