

# (Combinatorial) Characterization of Exponential Families of Lumpable Stochastic Matrices

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Dec. 18<sup>th</sup>, 2025

MaxEnt 2025 — University of Auckland, 14–19 Dec. 2025

## 1. Exponential Families of Stochastic Matrices

irreducible Markov chains,  $\mathfrak{s}$ -normalization, maximum entropy principle

## 2. Lumpings & Embeddings

definition, characterization, operational interpretation

## 3. Geometry of Lumpability

canonical embedding, foliation of the lumpable family, characterization of e-families



## Exponential Families of Stochastic Matrices

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- ◆ Finite space  $\mathcal{Y}$ . Probability distributions  $\mathcal{P}(\mathcal{Y})$ .
- ◆  $\mathcal{E} \subset \mathcal{Y}^2$  such that the digraph  $(\mathcal{Y}, \mathcal{E})$  is **strongly connected**.
- ◆ Functions and positive functions over  $\mathcal{E}$ :  $\mathcal{F}(\mathcal{Y}, \mathcal{E}), \quad \mathcal{F}^+(\mathcal{Y}, \mathcal{E})$ .
- ◆ **Irreducible** row-stochastic matrices over  $(\mathcal{Y}, \mathcal{E})$ :  $\mathcal{W}(\mathcal{Y}, \mathcal{E})$ .

Discrete-time, time-homogeneous Markov chain

$$\mathbb{P}(Y_t = y_t | Y_1 = y_1, \dots, Y_{t-1} = y_{t-1}) = \mathbb{P}(Y_t = y_t | Y_{t-1} = y_{t-1})$$

$$\mathbb{P}_\mu(Y_1 = y_1, \dots, Y_n = y_n) = \mu(y_1) \prod_{t=1}^{n-1} P(y_t, y_{t+1}),$$

$$(\mu, P) \in (\mathcal{P}(\mathcal{Y}), \mathcal{W}(\mathcal{Y}, \mathcal{E})).$$

- ◆ **Stationary distribution**:  $\pi P = \pi$ .
- ◆ **Edge-measure**:  $Q(y, y') = \pi(y)P(y, y') = \mathbb{P}_\pi(Y_t = y, Y_{t+1} = y')$ .

▷ **Probability distributions.**

$$u > 0, \quad \frac{u}{\|u\|_1}$$

▷ **Stochastic matrices.**

**Definition.  $\mathfrak{s}$ -normalization (Miller, 1961).**

When  $(\mathcal{Y}, \mathcal{E})$  is strongly connected we define the mapping

$$\begin{aligned} \mathfrak{s}: \mathcal{F}^+(\mathcal{Y}, \mathcal{E}) &\rightarrow \mathcal{W}(\mathcal{Y}, \mathcal{E}) \\ \mathfrak{s}(F)(y, y') &= \frac{F(y, y') v_F(y')}{\rho_F v_F(y)}, \end{aligned}$$

with  $\rho_F$  and  $v_F$  the Perron–Frobenius (PF) root and associated right eigenvector of  $F$ .

Polytope (m-family) generated by the set of linear constraints  $\{g_i = c_i\}$ ,

$$\mathcal{L} = \left\{ P \in \mathcal{W}(\mathcal{Y}, \mathcal{E}) : \sum_{(y, y') \in \mathcal{E}} Q(y, y') g_i(y, y') = c_i, \forall i \in [d] \right\} \subset \mathcal{W}(\mathcal{Y}, \mathcal{E}).$$

Let  $P \in \mathcal{W}(\mathcal{Y}, \mathcal{E})$ , and look at projection (information divergence rate  $D$ ) onto  $\mathcal{L}$ ,

$$P_e \triangleq \arg \min_{P' \in \mathcal{L}} D(P' || P).$$

Minimizer  $P_e$  belongs to an “exponential family”. For  $\lambda \in \mathbb{R}^d$ ,

$$\tilde{P}_\lambda(y, y') = P(y, y') \exp \left( \sum_{i \in [d]} \lambda^i g_i(y, y') \right),$$

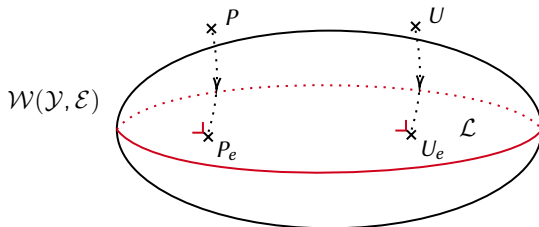
and for  $\psi(\lambda)$  log-PF root of  $\tilde{P}_\lambda$ ,  $P_e = \mathfrak{s}(\tilde{P}_{\lambda_*})$  with

$$\lambda^* = \arg \max_{\lambda \in \mathbb{R}^d} \{ \lambda \cdot c - \psi(\lambda) \}.$$

# Maximum Entropy Principle for Markov chains (Csiszár et al., 1987)

$$\text{Entropy rate: } H(P) \triangleq \lim_{k \rightarrow \infty} \frac{1}{k} H(Y_1, Y_2, \dots, Y_k)$$

$$U = \mathfrak{s}(\delta_{\mathcal{E}}), \text{ maxentropic in } \mathcal{W}(\mathcal{Y}, \mathcal{E}), \quad U_e \triangleq \arg \min_{P' \in \mathcal{L}} D(P' || U).$$



$$U_e = \arg \min_{P' \in \mathcal{L}} \left\{ -H(P') - \overbrace{\mathbb{E}_{(Y, Y') \sim Q'} [\log U(Y, Y')]}^{-\log \rho(\mathfrak{s}(\delta_{\mathcal{E}}))} \right\} = \arg \max_{P' \in \mathcal{L}} H(P').$$

## Exponential family (e-family) of stochastic matrices (Nagaoka, 2005)

$$\mathcal{V}_e = \left\{ P_\theta : \theta = (\theta^1, \dots, \theta^d) \in \mathbb{R}^d \right\} \subset \mathcal{W}(\mathcal{Y}, \mathcal{E}),$$

is e-family with natural parameter  $\theta$  and dimension  $d$ , when there exist a function  $K \in \mathcal{F}(\mathcal{Y}, \mathcal{E})$ ,  $d$  linearly independent functions  $G_1, \dots, G_d \in \mathcal{G}(\mathcal{Y}, \mathcal{E})$ , and functions  $R \in \mathbb{R}^{\Theta \times \mathcal{Y}}$ ,  $\psi \in \mathbb{R}^\Theta$ , such that

$$\log P_\theta(y, y') = K(y, y') + \sum_{i=1}^d \theta^i G_i(y, y') + R(\theta, y') - R(\theta, y) - \psi(\theta),$$

$$\text{i.e.} \quad P_\theta = \mathfrak{s} \circ \exp \left( K + \sum_{i=1}^d \theta^i G_i \right),$$

where  $\mathcal{G}(\mathcal{Y}, \mathcal{E})$  is the quotient space

$$\mathcal{G}(\mathcal{Y}, \mathcal{E}) \triangleq \mathcal{F}(\mathcal{Y}, \mathcal{E}) / \mathcal{N}(\mathcal{Y}, \mathcal{E}),$$

$$\mathcal{N}(\mathcal{Y}, \mathcal{E}) \triangleq \left\{ N : \exists f, c, N(y, y') = f(y') - f(y) + c \right\}.$$

**Expectation parameter ( $\clubsuit$ ):**  $\eta_i(\theta) = \sum_{(y, y') \in \mathcal{E}} Q_\theta(y, y') G_i(y, y')$ .



# Exponential families of stochastic matrices

**Example (Nagaoka, 2005).**  $\mathcal{W}(\mathcal{Y}, \mathcal{E})$  forms an e-family of dimension  $|\mathcal{E}| - |\mathcal{Y}|$ .

**Example. Parametrization of  $\mathcal{W}(\mathcal{Y}, \mathcal{Y}^2)$  proposed by Ito and Amari (1988).**

With  $\mathcal{Y} \cong [m]$ , pick  $y_\star \in \mathcal{Y}$ , and write,

$$\begin{aligned}\log P(y, y') &= \sum_{i=1, i \neq y_\star}^m \log \frac{P(y_\star, i)P(i, y_\star)}{P(y_\star, y_\star)P(y_\star, y_\star)} \delta_i(y') \\ &\quad + \sum_{i=1, i \neq y_\star}^m \sum_{j=1, j \neq y_\star}^m \log \frac{P(i, j)P(y_\star, y_\star)}{P(y_\star, j)P(i, y_\star)} \delta_i(y) \delta_j(y') \\ &\quad + \log P(y, y_\star) - \log P(y', y_\star) + \log P(y_\star, y_\star).\end{aligned}$$

Basis is given by

$$\begin{aligned}g_i &= 1^\top \delta_i, & i \in [m], i \neq y_\star \\ g_{ij} &= \delta_i^\top \delta_j, & i, j \in [m], i, j \neq y_\star\end{aligned}$$

and the parameters are

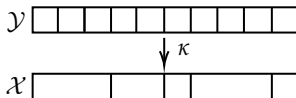
$$\theta^i = \log \frac{P(y_\star, i)P(i, y_\star)}{P(y_\star, y_\star)P(y_\star, y_\star)}, \quad \theta^{ij} = \log \frac{P(i, j)P(y_\star, y_\star)}{P(y_\star, j)P(i, y_\star)}.$$

## Lumpings & Embeddings

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- ◆  $\mathcal{Y}$  a **finite** alphabet.
  - ◆  $(\mathcal{Y}, \mathcal{E})$  be a **strongly connected digraph** with vertex set  $\mathcal{Y}$  and edge set  $\mathcal{E} \subset \mathcal{Y}^2$ .
  - ◆  $\mathcal{F}(\mathcal{Y}, \mathcal{E}) \cong \mathbb{R}^{\mathcal{E}}, \mathcal{F}^+(\mathcal{Y}, \mathcal{E}) \cong \mathbb{R}_+^{\mathcal{E}}$ .
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- ◆  $\mathcal{X}$  another **finite** alphabet with  $|\mathcal{X}| \leq |\mathcal{Y}|$ .
- ◆  $\kappa: \mathcal{Y} \rightarrow \mathcal{X}$  **surjective** map.
- ◆  $\mathcal{Y} = \bigsqcup_{x \in \mathcal{X}} \mathcal{S}_x$  where for  $x \in \mathcal{X}, \mathcal{S}_x \triangleq \kappa^{-1}(\{x\})$ .



- ◆  $\mathcal{D} \triangleq \kappa(\mathcal{E}) \triangleq \{(\kappa(y), \kappa(y')) : (y, y') \in \mathcal{E}\} \subset \mathcal{X}^2$ .
- ◆  $(\mathcal{Y}, \mathcal{E})$  strongly connected  $\implies (\mathcal{X}, \mathcal{D})$  strongly connected.

$\kappa$  maps the graph  $(\mathcal{Y}, \mathcal{E})$  to the graph  $(\mathcal{X}, \mathcal{D})$

Stationary Markov chain  $\{Y_t\}_{t \in \mathbb{N}} \sim P \in \mathcal{W}(\mathcal{Y}, \mathcal{E})$ .

**Data-processing:**

$$\kappa(Y_1), \kappa(Y_2), \kappa(Y_3), \dots$$

- ◆ State space compression (storage, interpretability, ...).
- ◆ Process  $\{\kappa(Y_t)\}_{t \in \mathbb{N}}$  is **not necessarily a Markov chain**,  
(Burke and Rosenblatt, 1958; Rogers and Pitman, 1981).
- ◆ Whenever Markovian, we say that the MC (or  $P$ ) is  $\kappa$ -lumpable.

**Characterization of  $\kappa$ -lumpability (Kemeny and Snell, 1983).**

$P \in \mathcal{W}(\mathcal{Y}, \mathcal{E})$  is  $\kappa$ -lumpable if and only if when for any  $(x, x') \in \mathcal{D}$  and any  $y_1, y_2 \in \mathcal{S}_x$ ,

$$\sum_{y' \in \mathcal{S}_{x'}} P(y_1, y') = \sum_{y' \in \mathcal{S}_{x'}} P(y_2, y').$$

Lumpable family:  $\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$

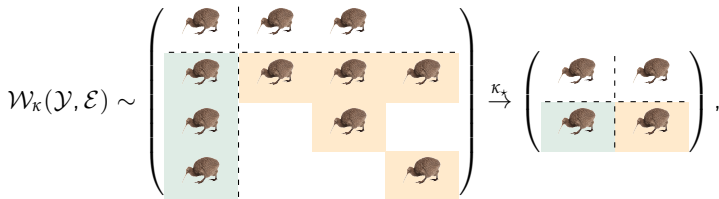
$$\kappa_\star: \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) \rightarrow \mathcal{W}(\mathcal{X}, \mathcal{D})$$

When  $P$  is  $\kappa$ -lumpable,

$$\kappa_\star P(x, x') \triangleq P(y, \mathcal{S}_{x'}), y \in \mathcal{S}_x.$$

**Visual representation upon relabeling (support, blocks and rows).**

$$\mathcal{Y} = \{0, 1, 2, 3\}, \mathcal{X} = \{a, b\}, \mathcal{S}_a = \{0\}, \mathcal{S}_b = \{1, 2, 3\}.$$



Let  $\kappa: \mathcal{Y} \rightarrow \mathcal{X}$  inducing the partition  $\bigsqcup_{x \in \mathcal{X}} \mathcal{S}_x = \mathcal{Y}$ .

**Definition.**  $\kappa$ -compatible Markov embedding.

$$\Lambda_\star: \mathcal{W}(\mathcal{X}, \mathcal{D}) \rightarrow \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$$

$$\text{where } \Lambda_\star P(y, y') = P(\kappa(y), \kappa(y')) \Lambda(y, y'), \forall (y, y') \in \mathcal{E},$$

$$\blacklozenge \Lambda \in \mathcal{F}^+(\mathcal{Y}, \mathcal{E}).$$

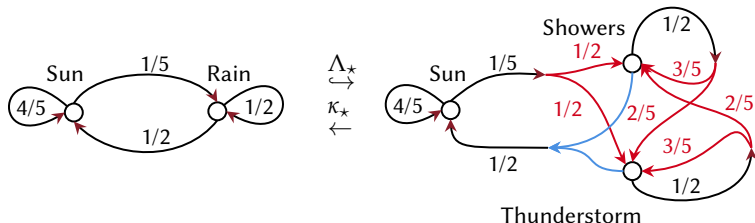
$$\blacklozenge \forall y \in \mathcal{Y}, x' \in \mathcal{X}, (\kappa(y), x') \in \mathcal{D} \implies (\Lambda(y, y'))_{y' \in \mathcal{S}_{x'}} \in \mathcal{P}(\mathcal{S}_{x'}).$$

$$\Lambda = \begin{pmatrix} w_{1,1} & w_{1,2} & \cdots & & w_{1,|\mathcal{X}|} \\ \vdots & & & & \vdots \\ w_{x,1} & \cdots & w_{x,x'} & \cdots & w_{x,|\mathcal{X}|} \\ \vdots & & & & \vdots \\ w_{|\mathcal{X}|,1} & & \cdots & & w_{|\mathcal{X}|,|\mathcal{X}|} \end{pmatrix}.$$

Can verify:  $\Lambda_\star P \in \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$ ,

$\kappa_\star \Lambda_\star P = P$  (right-inverse).

## Example: weather model



$$P = \begin{pmatrix} 4/5 & 1/5 \\ 1/2 & 1/2 \end{pmatrix},$$

$$\Lambda = \begin{pmatrix} 1 & 1/2 & 1/2 \\ 1 & 3/5 & 2/5 \\ 1 & 2/5 & 3/5 \end{pmatrix},$$

$$\Lambda_\star P = \begin{pmatrix} 4/5 & 1/10 & 1/10 \\ 1/2 & 3/10 & 1/5 \\ 1/2 & 1/5 & 3/10 \end{pmatrix}.$$

## Geometry of Lumpability

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## Classification of established families — Markov-centric properties

Manifold	m-family	e-family	Dimension	Reference
$\mathcal{W}(\mathcal{Y}, \mathcal{E})$	○	○	$ \mathcal{E}  -  \mathcal{Y} $	Nagaoka (2005)
$\mathcal{W}_{\text{iid}}(\mathcal{Y}, \mathcal{Y}^2)$	×	○	$ \mathcal{Y}  - 1$	
$\mathcal{W}_{\text{bis}}(\mathcal{Y}, \mathcal{Y}^2)$	○	×	$( \mathcal{Y}  - 1)^2$	Hayashi and Watanabe (2016)
$\mathcal{W}_{\text{rev}}(\mathcal{Y}, \mathcal{E})$	○	○	$( \mathcal{E}  +  \ell(\mathcal{E}) )/2 - 1$	W. and Watanabe (2021)
$\mathcal{W}_{\text{sym}}(\mathcal{Y}, \mathcal{Y}^2)$	○	×	$ \mathcal{Y}  ( \mathcal{Y}  - 1)/2$	
$\mathcal{W}_{\kappa}(\mathcal{Y}, \mathcal{E})$	?	?	?	W. and Watanabe (2024) Watanabe and W. (2024)

### Observation.

$\mathcal{W}_{\kappa}(\mathcal{Y}, \mathcal{E})$  is generally neither an e-family, nor an m-family.

...but we can decompose the family into simpler structures.

# Linear family of stochastic matrices that lump into prescribed $\bar{P}_0$

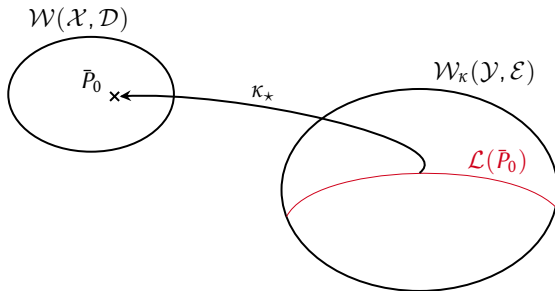
Let  $\bar{P}_0 \in \mathcal{W}(\mathcal{X}, \mathcal{D})$ , and

$$\mathcal{L}(\bar{P}_0) \triangleq \{P \in \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) : \kappa_\star P = \bar{P}_0\},$$

**Lemma.**

$\mathcal{L}(\bar{P}_0)$  forms an **m-family** in  $\mathcal{W}(\mathcal{Y}, \mathcal{E})$ , with

$$\dim \mathcal{L}(\bar{P}_0) = |\mathcal{E}| - \sum_{(x, x') \in \mathcal{D}} |\mathcal{S}_x|.$$



# Exponential family of embedding at some prescribed origin $P_{\odot}$

## Definition. Canonical embedding.

Let  $P \in \mathcal{W}_{\kappa}(\mathcal{Y}, \mathcal{E})$ . There exists a unique  $\Lambda_{\star}^{(P)}: \mathcal{W}(\mathcal{X}, \mathcal{D}) \rightarrow \mathcal{W}_{\kappa}(\mathcal{Y}, \mathcal{E})$  satisfying

$$P = \Lambda_{\star}^{(P)} \kappa_{\star} P.$$

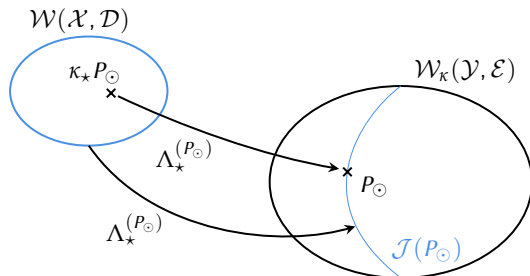
Let  $P_{\odot} \in \mathcal{W}_{\kappa}(\mathcal{Y}, \mathcal{E})$ , and  $\Lambda_{\star}^{(P_{\odot})}$  canonical embedding.

$$\mathcal{J}(P_{\odot}) \triangleq \left\{ \Lambda_{\star}^{(P_{\odot})} \bar{P} : \bar{P} \in \mathcal{W}(\mathcal{X}, \mathcal{D}) \right\} \subset \mathcal{W}_{\kappa}(\mathcal{Y}, \mathcal{E}).$$

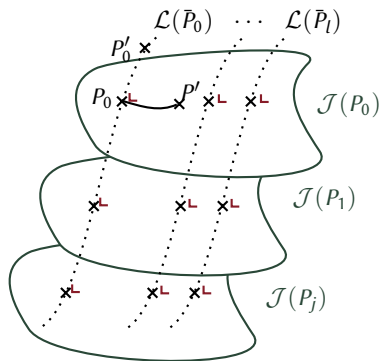
## Lemma.

$\mathcal{J}(P_{\odot})$  forms an e-family in  $\mathcal{W}(\mathcal{Y}, \mathcal{E})$ , with

$$\dim \mathcal{J}(P_{\odot}) = |\mathcal{D}| - |\mathcal{X}|.$$



# Foliated manifold of lumpable stochastic matrices



## Theorem.

For any fixed  $\bar{P}_0 \in \mathcal{W}(\mathcal{X}, \mathcal{D})$ ,

$$\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) = \biguplus_{P \in \mathcal{L}(\bar{P}_0)} \mathcal{J}(P).$$

$$\dim \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) = |\mathcal{E}| - \sum_{(x, x') \in \mathcal{D}} |\mathcal{S}_x| + |\mathcal{D}| - |\mathcal{X}|.$$

## Characterization problem.

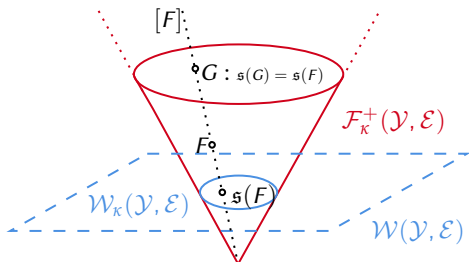
**When does  $\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$  form an e-family?**

single leaf only? conditions on  $\kappa$  and  $(\mathcal{Y}, \mathcal{E})$ ? algorithmic considerations?

▷ To analyze the properties of  $\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$ , we look at the relaxation  $\mathcal{F}_\kappa^+(\mathcal{Y}, \mathcal{E})$ .

$$\mathcal{F}_\kappa(\mathcal{Y}, \mathcal{E}) \triangleq \{\text{lumpable functions (not nec. stochastic)}\},$$

$$\mathcal{F}_\kappa^+(\mathcal{Y}, \mathcal{E}) \triangleq \{\text{positive lumpable cone}\}.$$



## Commutativity.

$\mathfrak{s}$ -normalization preserves  $\kappa$ -lumpability and

$$\begin{array}{ccc} \mathcal{F}_\kappa^+(\mathcal{Y}, \mathcal{E}) & \xrightarrow{\kappa_*} & \mathcal{F}^+(\mathcal{X}, \mathcal{D}) \\ \mathfrak{s} \downarrow & & \downarrow \mathfrak{s} \\ \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) & \xrightarrow{\kappa_*} & \mathcal{W}(\mathcal{X}, \mathcal{D}). \end{array}$$

# Merging blocks

▷ To analyze the properties of  $\mathcal{F}_k^+(\mathcal{Y}, \mathcal{E})$ , we will look at the structure of the blocks.

▷ Merging rows in block  $(x, x') \in \mathcal{D}$ :

$$\mathcal{M}_{x,x'} \triangleq \{y \in \mathcal{S}_x : \exists y'_1, y'_2 \in \mathcal{S}_{x'}, y'_1 \neq y'_2, (y, y'_1), (y, y'_2) \in \mathcal{E}\}.$$

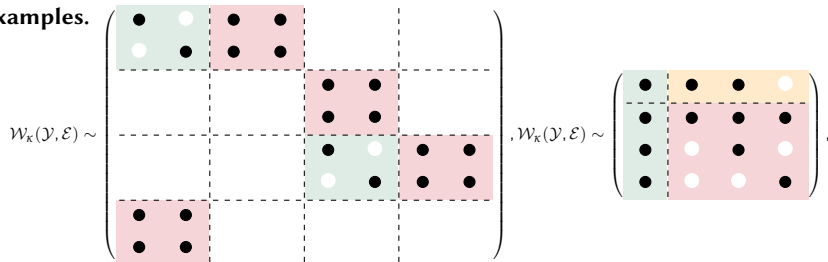
▷ Merging blocks:

$$\mathcal{M} \triangleq \{(x, x') \in \mathcal{D} : \mathcal{M}_{x,x'} \neq \emptyset\}.$$

▷ Multi-row merging blocks:

$$\mathcal{M}_{>1} \triangleq \{(x, x') \in \mathcal{M}, |\mathcal{S}_x| > 1\}.$$

Examples.



## Definition.

$$\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E}) \triangleq \{\log F: F \in \mathcal{F}_\kappa^+(\mathcal{Y}, \mathcal{E})\}$$

## Theorem. Characterization of log-affinity of the cone $\mathcal{F}_\kappa^+(\mathcal{Y}, \mathcal{E})$ .

The two following statements are equivalent.

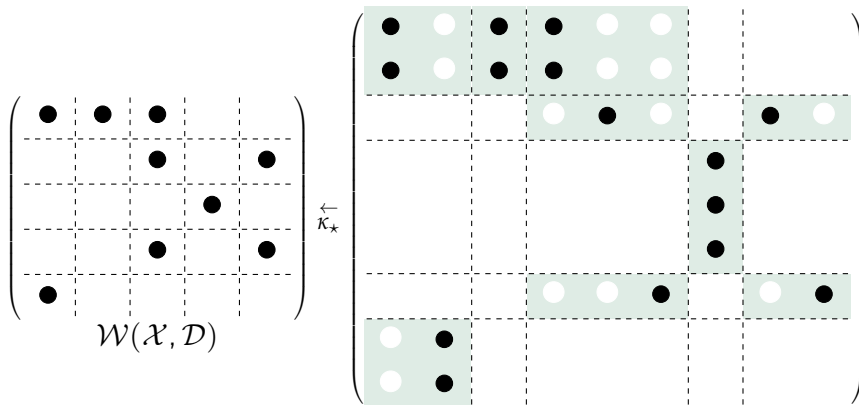
- (i)  $(\mathcal{Y}, \mathcal{E})$  has **no multi-row merging block** with respect to  $\kappa$ .
- (ii)  $\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E})$  is an **affine space**.

## Corollary. No multi-row merging block criterion.

If  $(\mathcal{Y}, \mathcal{E})$  has **no multi-row merging block**, then  $\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$  is **e-family**.

# Example: Hudson expansion (Kemeny and Snell, 1983)

$$\{X_t\}_t \sim P \in \mathcal{W}(\mathcal{X}, \mathcal{D}) \quad \{(X_t, X_{t+1})\}_t \sim ?$$



$$H_\star \mathcal{W}(\mathcal{X}, \mathcal{D}) = \mathcal{W}_\kappa(\mathcal{Y} = \mathcal{D}, \mathcal{E})$$

$$\mathcal{E} = \{(e = (x_1, x_2), e' = (x'_1, x'_2)) \in \mathcal{D}^2 : x_2 = x'_1\}, \kappa: \mathcal{Y} \rightarrow \mathcal{X}, (x_1, x_2) \mapsto x_2$$

**No merging row  $\implies \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$  forms an e-family.**



## Theorem. Monotonicity.

Let  $\mathcal{E} \subset \mathcal{E}'$ . If  $\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}')$  forms an e-family, then  $\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$  forms an e-family.

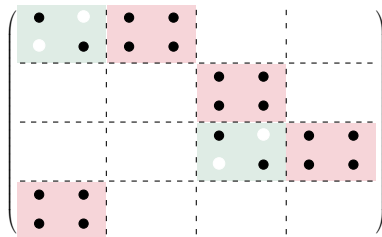
What happens if there is a multi-row merging block?

## Theorem. Lazy cycle criterion.

If for any  $P \in \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$ ,

$$P = D + \Pi,$$

$\kappa_\star \Pi$  is a permutation matrix over  $\mathcal{X}$  and  $D$  is diagonal, then  $\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$  forms an e-family.



- ◆ There can be **e-families with multi-row merging blocks**.
- ◆ There can be e-families with an **arbitrarily large number** of multi-row merging blocks.

## Redundant merging block criterion

**Theorem. Redundant merging block criterion.**

If  $(\mathcal{Y}, \mathcal{E})$  has a **redundant multi-row merging block**, then  $\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$  is **not e-family**.

$$\mathcal{Y} = \{0, 1, 2, 3, 4, 5\}, \mathcal{X} = \{a, b, c\}, \mathcal{S}_a = \{0, 1\}, \mathcal{S}_b = \{2, 3\}, \mathcal{S}_c = \{4, 5\},$$

$$\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) \sim \left( \begin{array}{cc|cc|cc} \bullet & \circ & \bullet & \bullet & \bullet & \circ \\ \circ & \bullet & \bullet & \bullet & \circ & \bullet \\ \hline \bullet & \circ & \bullet & \circ & \bullet & \bullet \\ \circ & \bullet & \circ & \bullet & \bullet & \bullet \\ \hline \bullet & \bullet & \bullet & \circ & \bullet & \circ \\ \bullet & \bullet & \circ & \bullet & \circ & \bullet \end{array} \right).$$

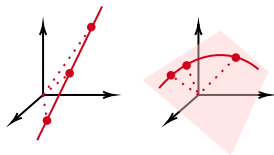
Remove merging block  $(b, c)$ . Closed path exhausting  $\mathcal{Y}$ :

$$1 \rightarrow 2 \rightarrow 0 \rightarrow 4 \rightarrow 2 \rightarrow 0 \rightarrow 3 \rightarrow 1 \rightarrow 5 \rightarrow 3 \rightarrow 1,$$

$$\implies \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) \text{ not e-family.}$$

▷ **Recall class:**

$$\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E}) \triangleq \{\log F : F \in \mathcal{F}_\kappa^+(\mathcal{Y}, \mathcal{E})\}.$$



▷ **Affine hull in  $\mathcal{F}(\mathcal{Y}, \mathcal{E})$ :**

$$\text{aff}(\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E})) \triangleq \left\{ \sum_{i=1}^k \alpha_i G_i : k \in \mathbb{N}, \alpha \in \mathbb{R}^k, \sum_{i=1}^k \alpha_i = 1, G_1, \dots, G_k \in \mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E}) \right\}.$$

▷ **Recall:**  $\text{aff}(\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E})) = \mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E})$  iff no multi-row merging block.

**Lemma. Linear space, basis and dimension.**

$$0_{\mathcal{F}(\mathcal{Y}, \mathcal{E})} \in \text{aff}(\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E})).$$

$$\dim \text{aff}(\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E})) = \text{Can compute.}$$

$$\text{Can construct a basis for } \text{aff}(\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E})).$$

## Theorem. Dimensional criterion.

$\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$  forms an **e-family** if and only if

$$\dim \left( \text{aff}(\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E})) \oplus \mathcal{N}(\mathcal{Y}, \mathcal{E}) \right) = |\mathcal{Y}| + |\mathcal{E}| + \underbrace{|\mathcal{D}| - |\mathcal{X}| - \sum_{(x, x') \in \mathcal{D}} |\mathcal{S}_x|}_{= \dim \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})}.$$

## Algorithm.

- ◆ 1. Construct a basis for  $\text{aff}(\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E}))$  (skipped in this talk).
- ◆ 2. Construct a basis for  $\mathcal{N}(\mathcal{Y}, \mathcal{E}) = \{N: N(y, y') = f(y') - f(y) + c\}$ .
- ◆ 3. Concatenate the two bases for  $\text{aff}(\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E}))$  and  $\mathcal{N}(\mathcal{Y}, \mathcal{E})$ .
- ◆ 4. Compute the rank of the system (flatten into vectors of dimension  $|\mathcal{E}|$ ).

## Dimensional criterion – Example

$$\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) \sim \left( \begin{array}{ccc|ccc} \bullet & & & \bullet & & \bullet \\ \bullet & & & \bullet & & \bullet \\ \bullet & & & & \bullet & \bullet \\ \bullet & & & & \bullet & \bullet \end{array} \right),$$

▷ The only multi-row merging block is **not redundant**.

▷ Basis for  $\text{aff}(\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E}))$ .

$$\begin{pmatrix} 1 & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \end{pmatrix}, \begin{pmatrix} & 1 & & & & \\ & & & & & \\ & & & & & \\ & & & & & \end{pmatrix}, \begin{pmatrix} & & 1 & & & \\ & & & & & \\ & & & & & \\ & & & & & \end{pmatrix}, \begin{pmatrix} & & & 1 & & \\ & & & 1 & & \\ & & & 1 & & \\ & & & & & \end{pmatrix},$$

$$\begin{pmatrix} & & & & 1 & \\ & & & & & \\ & & & & & \\ & & & & & \end{pmatrix}, \begin{pmatrix} & & 1 & & & \\ & & & & & \\ & & & & & \\ & & & & & \end{pmatrix}, \begin{pmatrix} & & & 1 & & \\ & & & & & \\ & & & & & \\ & & & & & \end{pmatrix}, \begin{pmatrix} & & & & & 1 \\ & & & & & \\ & & & & & \\ & & & & & \end{pmatrix}.$$

## Dimensional criterion – Example

$$\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) \sim \left( \begin{array}{ccc|ccc} \bullet & & & \bullet & & \bullet \\ \bullet & & & \bullet & & \\ \bullet & & & & \bullet & \bullet \\ \bullet & & & & \bullet & \bullet \end{array} \right),$$

▷ Basis for  $\mathcal{N}(\mathcal{Y}, \mathcal{E})$

$$\left( \begin{array}{c|cc} 1 & 1 & 1 \\ \hline 1 & 1 & 1 \\ & 1 & \\ 1 & & 1 \\ & & 1 \end{array} \right), \left( \begin{array}{c|c} 1 & \\ \hline -1 & -1 \end{array} \right), \left( \begin{array}{c|c} 1 & \\ \hline -1 & 1 \end{array} \right), \left( \begin{array}{c|c} 1 & \\ \hline -1 & \end{array} \right).$$

$$\dim \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) = |\mathcal{E}| + |\mathcal{D}| - |\mathcal{X}| - \sum_{(x, x') \in \mathcal{D}} |\mathcal{S}_x| = 11 + 4 - 2 - 8 = 5$$

$$\dim(\mathcal{N}(\mathcal{Y}, \mathcal{E})) = |\mathcal{Y}| = 4$$

$$\dim(\text{aff}(\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E})) \oplus \mathcal{N}(\mathcal{Y}, \mathcal{E})) \stackrel{\text{comp.}}{=} 10 > 4 + 5.$$

$\Rightarrow \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$  not e-family.

Find Embedding

Lazy Cycle Criterion

$$\mathcal{O}(|\mathcal{X}|^2 + |\mathcal{E}|)$$

No Merging Block Criterion

$$\mathcal{O}(|\mathcal{X}| |\mathcal{Y}| + |\mathcal{E}|)$$

Dimension  
Criterion

$$\mathcal{O}(|\mathcal{E}|^\omega)$$

Witness  
Verification

$$\mathcal{O}(|\mathcal{Y}|^\omega)$$

$\Rightarrow$  Corollary  
 $\mathcal{O}(|\mathcal{X}| |\mathcal{Y}| + |\mathcal{E}|)$

Redundant Block  
Criterion  
 $\mathcal{O}(|\mathcal{D}| (|\mathcal{Y}| + |\mathcal{E}|))$

Monotonicity

Thank you for listening.



arXiv:2412.08400

◆ **Characterization of Exponential Families of Lumpable Stochastic Matrices**

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## References

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