

(Combinatorial) Characterization of Exponential Families of Lumpable Stochastic Matrices

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Outline

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

Irreducible Markov chains

- ◆ 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} : $\mathcal{F}(\mathcal{Y}, \mathcal{E})$, $\mathcal{F}^+(\mathcal{Y}, \mathcal{E})$.
- ◆ **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**: $Q(y, y') = \pi(y)P(y, y') = \mathbb{P}_\pi(Y_t = y, Y_{t+1} = y')$.

\mathfrak{s} -normalization

- ▷ 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

$$\mathfrak{s}: \mathcal{F}^+(\mathcal{Y}, \mathcal{E}) \rightarrow \mathcal{W}(\mathcal{Y}, \mathcal{E})$$

$$\mathfrak{s}(F)(y, y') = \frac{F(y, y')v_F(y')}{\rho_F v_F(y)},$$

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

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

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} ,

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

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

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

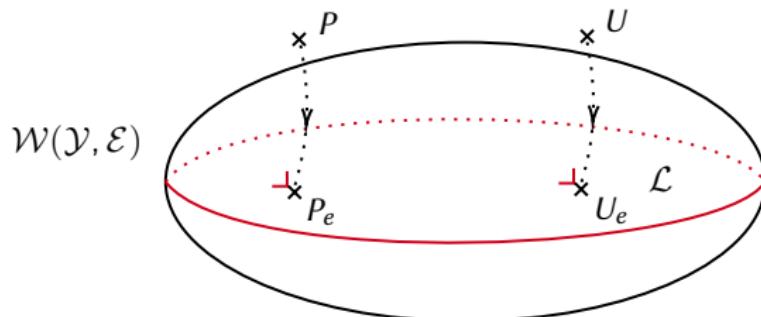
and for $\psi(\lambda)$ log-PF root of \tilde{P}_λ , $P_e = \mathfrak{s}(\tilde{P}_{\lambda_*})$ with

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

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

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

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



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

Exponential families of stochastic matrices

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 θ 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

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

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

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

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,

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

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

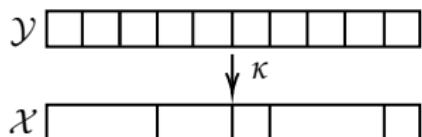
$$\theta^i = \log \frac{P(y_\star, i)P(i, y_\star)}{P(y_\star, y_\star)P(y_\star, y_\star)}, \quad \theta^{ij} = \log \frac{P(i, j)P(y_\star, y_\star)}{P(y_\star, j)P(i, y_\star)}.$$

Lumpings & Embeddings

Notation & Lumping map

- ◆ \mathcal{Y} a **finite** alphabet.
 - ◆ $(\mathcal{Y}, \mathcal{E})$ be a **strongly connected digraph** with vertex set \mathcal{Y} and edge set $\mathcal{E} \subset \mathcal{Y}^2$.
 - ◆ $\mathcal{F}(\mathcal{Y}, \mathcal{E}) \cong \mathbb{R}^{\mathcal{E}}, \mathcal{F}^+(\mathcal{Y}, \mathcal{E}) \cong \mathbb{R}_+^{\mathcal{E}}$.
-

- ◆ \mathcal{X} another **finite** alphabet with $|\mathcal{X}| \leq |\mathcal{Y}|$.
- ◆ $\kappa: \mathcal{Y} \rightarrow \mathcal{X}$ **surjective** map.
- ◆ $\mathcal{Y} = \biguplus_{x \in \mathcal{X}} \mathcal{S}_x$ 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.

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

Lumpability

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 κ -lumpable.

Characterization of κ -lumpability (Kemeny and Snell, 1983).

$P \in \mathcal{W}(\mathcal{Y}, \mathcal{E})$ is κ -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

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

When P is κ -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).

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

$$\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) \sim \left(\begin{array}{|c|c|c|c|} \hline & \text{kiwi} & \text{kiwi} & \text{kiwi} \\ \hline \text{kiwi} & \text{kiwi} & \text{kiwi} & \text{kiwi} \\ \hline \text{kiwi} & & \text{kiwi} & \\ \hline \text{kiwi} & & & \text{kiwi} \\ \hline \end{array} \right) \xrightarrow{\kappa_\star} \left(\begin{array}{|c|c|} \hline & \text{kiwi} \\ \hline \text{kiwi} & \text{kiwi} \\ \hline \end{array} \right),$$

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

Definition. κ -compatible Markov embedding.

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

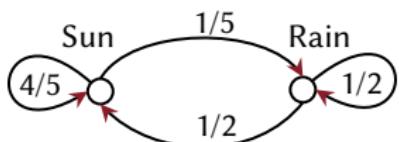
$$\text{where } \Lambda_* P(y, y') = P(\kappa(y), \kappa(y')) \Lambda(y, y'), \forall (y, y') \in \mathcal{E},$$

- ◆ $\Lambda \in \mathcal{F}^+(\mathcal{Y}, \mathcal{E})$.
- ◆ $\forall y \in \mathcal{Y}, x' \in \mathcal{X}, (\kappa(y), x') \in \mathcal{D} \implies (\Lambda(y, y'))_{y' \in \mathcal{S}_{x'}} \in \mathcal{P}(\mathcal{S}_{x'})$.

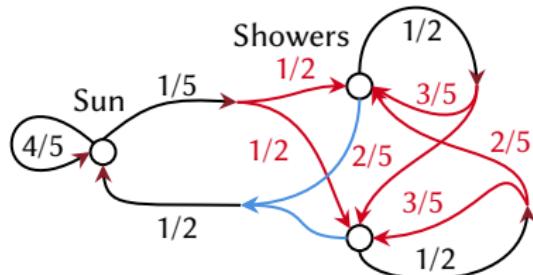
$$\Lambda = \begin{pmatrix} W_{1,1} & W_{1,2} & \cdots & & W_{1,|\mathcal{X}|} \\ \vdots & & & & \vdots \\ W_{x,1} & \cdots & W_{x,x'} & \cdots & W_{x,|\mathcal{X}|} \\ \vdots & & & & \vdots \\ W_{|\mathcal{X}|,1} & & \cdots & & W_{|\mathcal{X}|,|\mathcal{X}|} \end{pmatrix}.$$

Can verify: $\Lambda_* P \in \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$,
 $\kappa_* \Lambda_* P = P$ (right-inverse).

Example: weather model



$$\begin{matrix} \Lambda_\star \\ \xrightarrow{\quad} \\ \kappa_\star \\ \leftarrow \end{matrix}$$



$$P = \begin{pmatrix} 4/5 & 1/5 \\ 1/2 & 1/2 \end{pmatrix}, \quad \Lambda = \begin{pmatrix} 1 & 1/2 & 1/2 \\ 1 & 3/5 & 2/5 \\ 1 & 2/5 & 3/5 \end{pmatrix},$$

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

Geometry of Lumpability

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

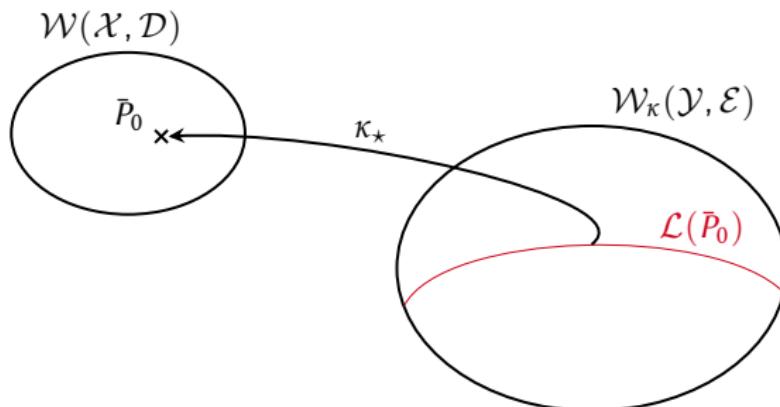
Let $\bar{P}_0 \in \mathcal{W}(\mathcal{X}, \mathcal{D})$, and

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

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

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

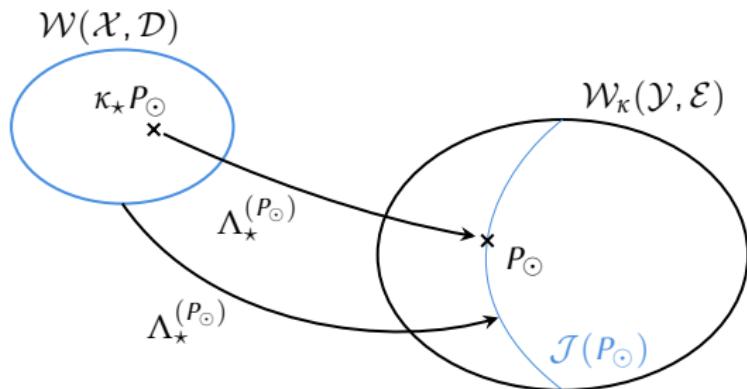
Let $P_\odot \in \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E})$, and $\Lambda_\star^{(P_\odot)}$ canonical embedding.

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

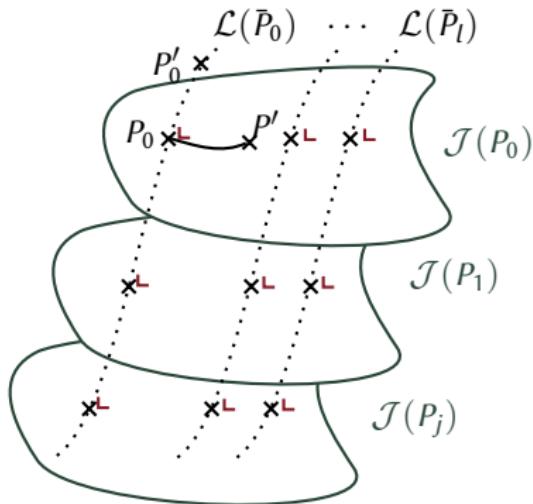
Lemma.

$\mathcal{J}(P_\odot)$ 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 $\bar{P}_0 \in \mathcal{W}(\mathcal{X}, \mathcal{D})$,

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

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

Characterization problem.

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

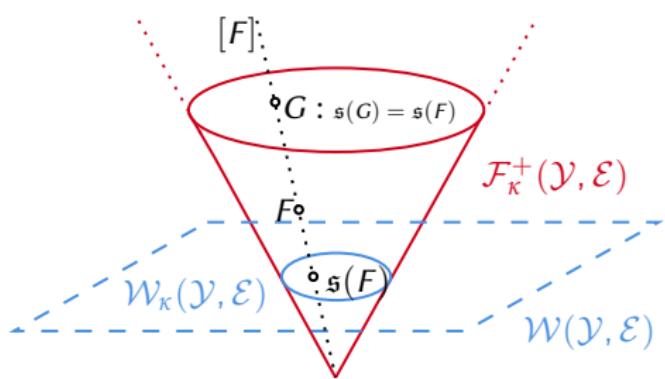
single leaf only? conditions on κ and $(\mathcal{Y}, \mathcal{E})$? algorithmic considerations?

Lumpable cone

▷ 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.

s -normalization preserves κ -lumpability and

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

Merging blocks

- ▷ To analyze the properties of $\mathcal{F}_\kappa^+(\mathcal{Y}, \mathcal{E})$, 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.

$$\mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) \sim \left(\begin{array}{c|c|c|c} \text{●} & \text{●} & & \\ \text{●} & & & \\ & \text{●} & \text{●} & \\ \hline & \text{●} & \text{●} & \\ & \text{●} & \text{●} & \\ \hline & & & \\ \hline & \text{●} & \text{●} & \\ & \text{●} & \text{●} & \\ \hline \end{array} \right), \quad \mathcal{W}_\kappa(\mathcal{Y}, \mathcal{E}) \sim \left(\begin{array}{c|c|c|c} \text{●} & \text{●} & \text{●} & \text{●} \\ \text{●} & & & \\ \text{●} & \text{●} & \text{●} & \\ \hline & \text{●} & \text{●} & \\ & \text{●} & \text{●} & \\ \hline & & & \\ \hline & \text{●} & \text{●} & \\ & \text{●} & \text{●} & \\ \hline \end{array} \right).$$

Lumpable cone

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.

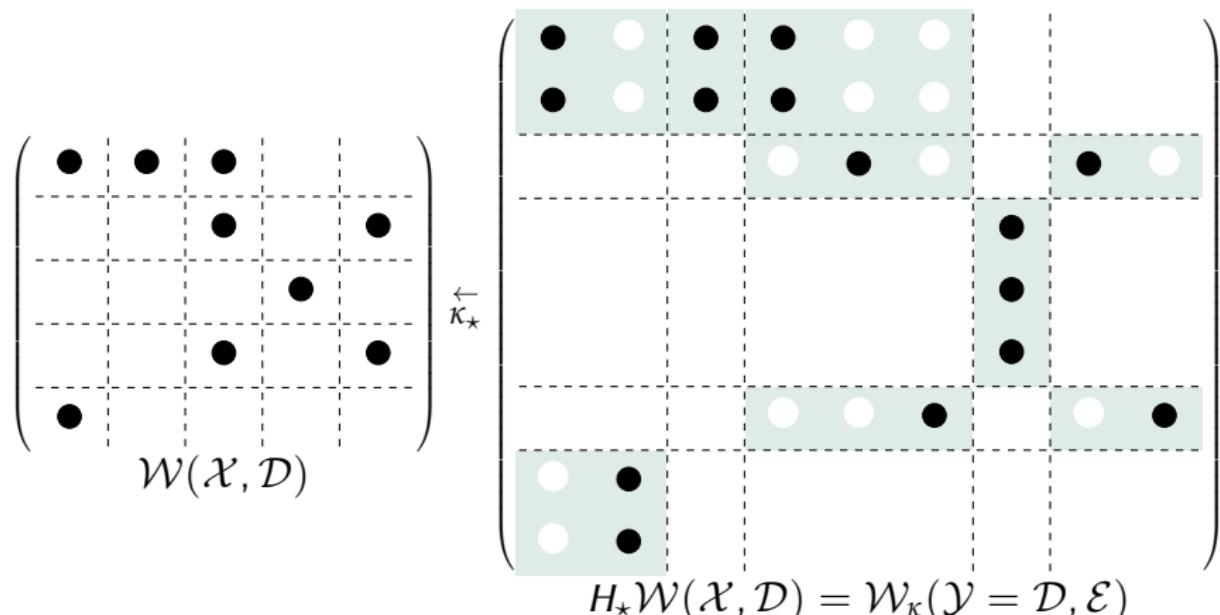
- (i) $(\mathcal{Y}, \mathcal{E})$ has **no multi-row merging block** with respect to κ .
- (ii) $\mathcal{G}_\kappa(\mathcal{Y}, \mathcal{E})$ is an **affine space**.

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)

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



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

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

Monotonicity & Lazy cycle criterion

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.

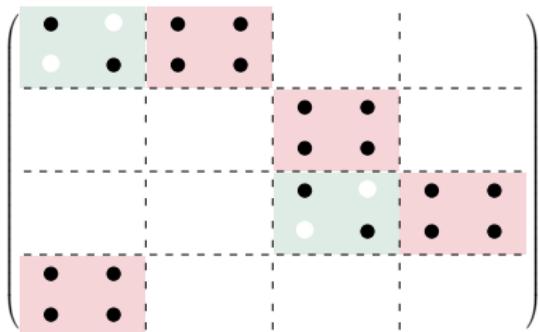
What happens if there is a multi-row merging block?

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 $\mathcal{W}_k(\mathcal{Y}, \mathcal{E})$ is **not e-family**.

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

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

$\mathcal{W}_k(\mathcal{Y}, \mathcal{E}) \sim$

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

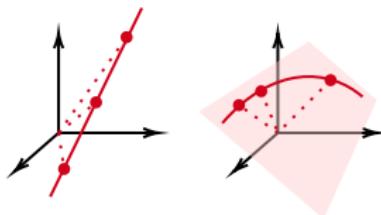
$$1 \rightarrow 2 \rightarrow 0 \rightarrow 4 \rightarrow 2 \rightarrow 0 \rightarrow 3 \rightarrow 1 \rightarrow 5 \rightarrow 3 \rightarrow 1,$$

$\implies \mathcal{W}_k(\mathcal{Y}, \mathcal{E})$ **not e-family**.

Dimensional criterion

▷ 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.

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

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

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

Dimensional criterion

Theorem. Dimensional criterion.

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

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

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

$$\mathcal{W}_k(\mathcal{Y}, \mathcal{E}) \sim \begin{pmatrix} \bullet & \bullet & \bullet \\ \bullet & \bullet & \bullet \end{pmatrix},$$

- ▷ The only multi-row merging block is **not redundant**.
- ▷ Basis for $\text{aff}(\mathcal{G}_k(\mathcal{Y}, \mathcal{E}))$.

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

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

Dimensional criterion – Example

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

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

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

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

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

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

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

Conclusion

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

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

\implies 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.



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

S. Watanabe, G. Wolfer

arXiv:2412.08400

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