

Pathwise duality for monotone Markov processes.

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joint with A. Sturm and N. Latz (in progress)

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Probability kernels

For general sets S, T , let $\mathcal{F}(S, T)$ denote the set of all functions $f : S \rightarrow T$.

A *random mapping representation* of a probability kernel K from S to T is an $\mathcal{F}(S, T)$ -valued random variable M such that

$$K(x, y) = \mathbb{P}[M(x) = y] \quad (x \in S, y \in T).$$

We say that K is *representable* in $\mathcal{G} \subset \mathcal{F}(S, T)$ if M can be chosen so that it takes values in \mathcal{G} . We set

$$Kf(x) := \sum_{y \in T} K(x, y)f(y) = \mathbb{E}[f(M(x))]$$
$$(x \in S, f \in \mathcal{F}(T, \mathbb{R})).$$

Monotone probability kernels

For partially ordered sets S, T , let $\mathcal{F}_{\text{mon}}(S, T)$ be the set of all monotone maps $m : S \rightarrow T$, i.e., those for which $x \leq x'$ implies $m(x) \leq m(x')$.

A probability kernel K is called *monotone* if

$$Kf \in \mathcal{F}_{\text{mon}}(S, \mathbb{R}) \quad \forall f \in \mathcal{F}_{\text{mon}}(T, \mathbb{R}),$$

and *monotonically representable* if K is representable in $\mathcal{F}_{\text{mon}}(S, T)$.

Monotonical representability implies monotonicity:

$$\begin{aligned} f \in \mathcal{F}_{\text{mon}}(T, \mathbb{R}) \quad \text{and} \quad x \leq x' &\Rightarrow \\ Kf(x) = \mathbb{E}[f(M(x))] &\leq \mathbb{E}[f(M(x'))] = Kf(x'). \end{aligned}$$

Monotone probability kernels

J.A. Fill & M. Machida (AOP 2001) (and also D.A. Ross (unpublished)) discovered that the converse does not hold. There are counterexamples with $S = T = \{0, 1\}^2$.

On the positive side, Kamae, Krengel & O'Brien (1977) and Fill & Machida (2001) have shown that:

(Sufficient conditions for monotone representability)

Let S, T be finite partially ordered sets and assume that at least one of the following conditions is satisfied:

- (i) *S is totally ordered.*
- (ii) *T is totally ordered.*

Then any monotone probability kernel from S to T is monotonically representable.

In particular, setting $S = \{1, 2\}$, this proves that if μ_1, μ_2 are probability laws on T such that

$$\mu_1 f \leq \mu_2 f \quad \forall f \in \mathcal{F}_{\text{mon}}(T, \mathbb{R}),$$

then it is possible to couple random variables M_1, M_2 with laws μ_1, μ_2 such that $M_1 \leq M_2$.

Markov semigroups

Let S be finite. By definition, a *Markov semigroup* is a collection of probability kernels $(P_t)_{t \geq 0}$ on S such that

$$P_0 = \lim_{t \downarrow 0} P_t = 1 \quad \text{and} \quad P_s P_t = P_{s+t}.$$

Each Markov semigroup is of the form

$$P_t := e^{tG} = \sum_{n=0}^{\infty} \frac{1}{n!} t^n G^n \quad (t \geq 0),$$

where the *generator* G satisfies

$$G(x, y) \geq 0 \quad (x \neq y) \quad \text{and} \quad \sum_{y \in S} G(x, y) = 0 \quad (x \in S).$$

We write

$$Gf(x) := \sum_{y \in T} G(x, y)f(y) \quad (x \in S, f \in \mathcal{F}(T, \mathbb{R})).$$

Representability of semigroups

By definition, G is *representable* in $\mathcal{G} \subset \mathcal{F}(S, S)$ if G can be written as

$$Gf(x) = \sum_{m \in \mathcal{G}} r_m (f(m(x)) - f(x)),$$

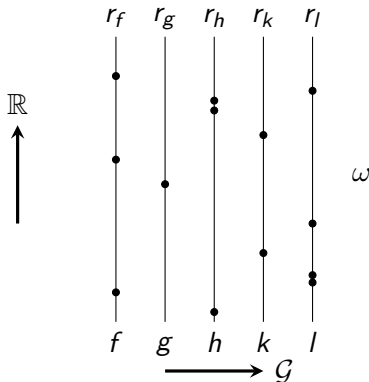
where $(r_m)_{m \in \mathcal{G}}$ are nonnegative constants (rates).

(Representability of semigroups)

Assume that \mathcal{G} is closed under composition and contains the identity map. Then the following statements are equivalent:

- (i) G can be represented in \mathcal{G} .
- (ii) P_t can be represented in \mathcal{G} for all $t \geq 0$.

Proof of (i) \Rightarrow (ii)

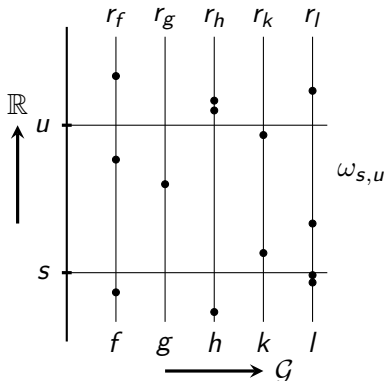


Let ρ be the measure on \mathcal{G} defined by $\rho(\{m\}) := r_m$.

Let ℓ denote the Lebesgue measure on \mathbb{R} .

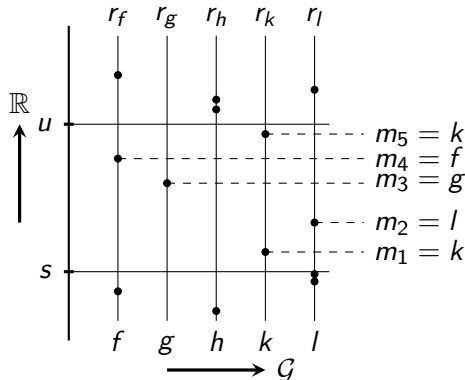
Let ω be a Poisson subset of $\mathcal{G} \times \mathbb{R}$ with intensity measure $\rho \otimes \ell$.

Proof of (i) \Rightarrow (ii)



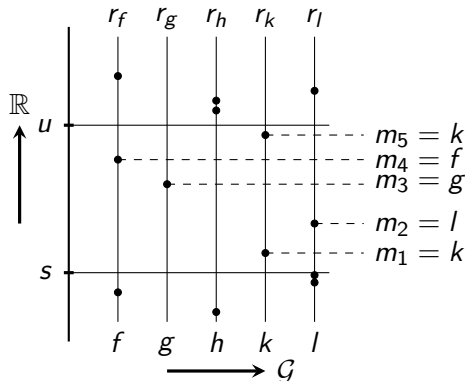
Let $\omega_{s,u} := \{(m, t) \in \omega : s < t \leq u\}$.

Proof of (i) \Rightarrow (ii)



Order the elements of $\omega_{s,u} := \{(m, t) \in \omega : s < t \leq u\}$
 as $\omega_{s,u} = \{(m_1, t_1), \dots, (m_n, t_n)\}$ with $t_1 < \dots < t_n$.

Proof of (i) \Rightarrow (ii)



Define $\mathbb{X}_{s,u} := m_n \circ \cdots \circ m_1$.

Stochastic flows

The random maps $(\mathbb{X}_{s,u})_{s \leq u}$ form a *stochastic flow*:

$$\mathbb{X}_{s,s} = 1 \quad \text{and} \quad \mathbb{X}_{t,u} \circ \mathbb{X}_{s,t} = \mathbb{X}_{s,u} \quad (s \leq t \leq u),$$

with independent increments:

$$\mathbb{X}_{t_0,t_1}, \dots, \mathbb{X}_{t_{n-1},t_n} \quad \text{independent for} \quad t_0 < \dots < t_n.$$

If X_0 is independent of ω , then

$$X_t := \mathbb{X}_{0,t}(X_0) \quad (t \geq 0)$$

defines a Markov process $(X_t)_{t \geq 0}$ with generator G , and

$$P_t(x, y) = \mathbb{P}[\mathbb{X}_{0,t}(x) = y]$$

gives the desired random mapping representation of the Markov semigroup $(P_t)_{t \geq 0}$ with generator G . ■

We call the Poisson set ω a *graphical representation* of X .

Note: Since $\omega_{s,u} := \{(m, t) \in \omega : s < t \leq u\}$,
the stochastic flow $\mathbb{X}_{s,t}$ is right-continuous in s and t .
As a result, $(X_t)_{t \geq 0}$ has right-continuous sample paths.

Setting $\omega_{s,u}^- := \{(m, t) \in \omega : s \leq t < u\}$
yields a stochastic flow $\mathbb{X}_{s,t}^-$ with left-continuous sample paths.

Two Markov processes X and Y with state spaces S and R are *dual* with *duality function* $\psi : S \times R \rightarrow \mathbb{R}$ iff

$$\mathbb{E}[\psi(X_t, Y_0)] = \mathbb{E}[\psi(X_0, Y_t)] \quad (*).$$

for all deterministic initial states X_0 and Y_0 .

If $(*)$ holds for deterministic initial states, then also for random initial states, provided X_t is independent of Y_0 and X_0 is independent of Y_t .

In terms of semigroups $(P_t)_{t \geq 0}$, $(Q_t)_{t \geq 0}$ and generators G, H , duality says

$$\begin{aligned} P_t \psi &= \psi Q_t^\dagger & (t \geq 0), \\ \Leftrightarrow \quad G \psi &= \psi H^\dagger, \end{aligned}$$

where A^\dagger denotes the transpose of a matrix A .

Pathwise duality

Two maps $m : S \rightarrow S$ and $\hat{m} : R \rightarrow R$ are *dual* w.r.t. a duality function $\psi : S \times R \rightarrow T$ iff

$$\psi(m(x), y) = \psi(x, \hat{m}(y)) \quad (x \in S, y \in R).$$

Assume that each $m \in \mathcal{G}$ has a dual map \hat{m} .

Let ω be a graphical representation for the process with generator

$$Gf(x) = \sum_{m \in \mathcal{G}} r_m(f(m(x)) - f(x)).$$

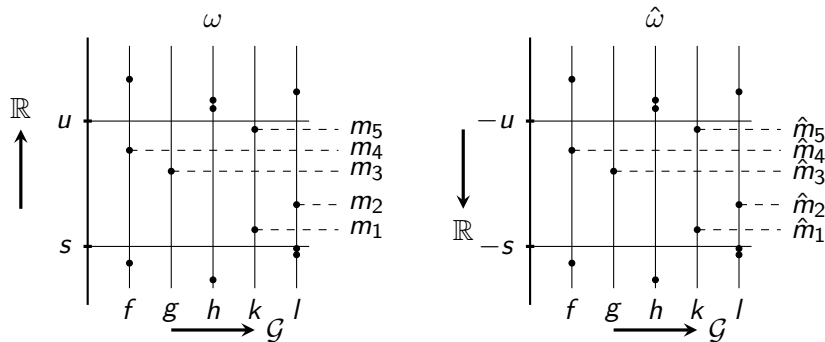
Then

$$\hat{\omega} := \{(\hat{m}, -t) : (m, t) \in \omega\}$$

is a graphical representation for the process with generator

$$Hf(y) = \sum_{m \in \mathcal{G}} r_m(f(\hat{m}(y)) - f(y)).$$

Pathwise duality



$$\mathbb{X}_{s,u}^- = m_1 \circ \cdots \circ m_1 \quad \text{and} \quad \mathbb{Y}_{-u,-s} = \hat{m}_1 \circ \cdots \hat{m}_5.$$

Pathwise duality

The stochastic flows $(\mathbb{X}_{s,u})_{s \leq u}$ constructed from ω and $(\mathbb{Y}_{s,u})_{s \leq u}$ constructed from $\hat{\omega}$ are dual:

$$\psi(\mathbb{X}_{s,u}^-(x), y) = \psi(x, \mathbb{Y}_{-u, -s}(y)) \quad (x \in S, y \in R),$$

where $\mathbb{X}_{s,u}^-$ denotes the left-continuous modification of $\mathbb{X}_{s,u}$.

Proof

$$\begin{aligned} & \psi(m_5 \circ m_4 \circ m_3 \circ m_2 \circ m_1(x), y) \\ &= \psi(m_4 \circ m_3 \circ m_2 \circ m_1(x), \hat{m}_5(y)) \\ &= \psi(m_3 \circ m_2 \circ m_1(x), \hat{m}_4 \circ \hat{m}_5(y)) \\ &= \dots = \psi(x, \hat{m}_1 \circ \hat{m}_2 \circ \hat{m}_3 \circ \hat{m}_4 \circ \hat{m}_5(y)). \end{aligned}$$



Pathwise duality

Two Markov processes X and Y are *pathwise dual* if they can be constructed from stochastic flows that are dual.

For duality functions $\psi : S \times R \rightarrow \mathbb{R}$, pathwise duality implies duality:

$$\begin{aligned}\mathbb{E}[\psi(X_t, Y_0)] &= \mathbb{E}[\psi(\mathbb{X}_{0,t}^-(X_0), Y_0)] \\ &= \mathbb{E}[\psi(X_0, \mathbb{Y}_{-t,0}(Y_0))] = \mathbb{E}[\psi(X_0, Y_t)].\end{aligned}$$

Even though pathwise duality is much stronger than duality, lots of well-known dualities can be realized as pathwise dualities.

(Pathwise duality) *If the generators G and H of X and Y have random mapping representations of the form*

$$\begin{aligned} Gf(x) &= \sum_{m \in \mathcal{G}} r_m(f(m(x)) - f(x)), \\ Hf(x) &= \sum_{m \in \mathcal{G}} r_m(f(\hat{m}(y)) - f(y)), \end{aligned}$$

where each map \hat{m} is a dual of m , then X and Y are pathwise dual.

A formal dual

Let $\mathcal{F}(S, T)$ be the set of functions $f : S \rightarrow T$.

Let $(\mathbb{X}_{s,u})_{s \leq u}$ be a stochastic flow on S . Then

$$\mathbb{F}_{s,u}(f) := f \circ \mathbb{X}_{-u,-s}^- \quad (f \in \mathcal{F}(S, T))$$

defines a stochastic flow $(\mathbb{F}_{s,u})_{s \leq u}$ on $\mathcal{F}(S, T)$.

If F_0 is an $\mathcal{F}(S, T)$ -valued random variable, independent of $(\mathbb{X}_{s,u})_{s \leq u}$, then

$$F_t := \mathbb{F}_{0,t}(F_0) = F_0 \circ \mathbb{X}_{-t,0}^- \quad (t \geq 0)$$

defines a Markov process $(F_t)_{t \geq 0}$ with values in $\mathcal{F}(S, T)$.

This Markov process is pathwise dual to X with duality function

$$\psi(x, f) := f(x) \quad (x \in S, f \in \mathcal{F}(S, T)).$$

Indeed

$$\psi(\mathbb{X}_{s,u}^-(x), f) = f \circ \mathbb{X}_{s,u}^-(x) = \mathbb{F}_{-u,-s}(f)(x) = \psi(x, \mathbb{F}_{-u,-s}(f)).$$

Invariant subspaces

Def A subspace $\mathcal{H} \subset \mathcal{F}(S, T)$ is *invariant* under the action of $(\mathbb{F}_{s,u})_{s \leq u}$ if $f \in \mathcal{H}$ implies $\mathbb{F}_{s,u}(f) \in \mathcal{H}$. That means

$$f \in \mathcal{H} \quad \Rightarrow \quad f \circ \mathbb{X}_{s,u} \in \mathcal{H} \quad (s \leq u).$$

Interesting pathwise duals are associated with *invariant subspaces* of $\mathcal{F}(S, T)$.

Assume that $\mathcal{H} \subset \mathcal{F}(S, T)$ is invariant and

$$R \ni y \mapsto \psi(\cdot, y) \in \mathcal{H}$$

is a bijection. Then setting

$$\mathbb{F}_{s,u}(\psi(\cdot, y)) =: \psi(\cdot, \mathbb{Y}_{s,u}(y))$$

defines a stochastic flow $(\mathbb{Y}_{s,u})_{s \leq u}$ on R that is dual to $(\mathbb{X}_{s,u})_{s \leq u}$ with duality function ψ .

A bit of order theory

Let S be a finite partially ordered space. The “upset” and “downset” of $A \subset S$ are defined as

$$A^\uparrow := \{x \in S : x \geq a \text{ for some } a \in A\},$$

$$A^\downarrow := \{x \in S : x \leq a \text{ for some } a \in A\}.$$

A set $A \subset S$ is *increasing* (resp. *decreasing*) if $A^\uparrow = A$ (resp. $A^\downarrow = A$).

A *lattice* is a partially ordered set such that for every $x, y \in S$ there exist $x \vee y \in S$ and $x \wedge y \in S$, called the *supremum* or *join* and *infimum* or *meet* of x and y , respectively, such that

$$\{x\}^\uparrow \cap \{y\}^\uparrow = \{x \vee y\}^\uparrow \quad \text{and} \quad \{x\}^\downarrow \cap \{y\}^\downarrow = \{x \wedge y\}^\downarrow.$$

A bit of order theory

Let S be a partially ordered set. A map $m : S \rightarrow S$ is *monotone* if

$$x \leq y \quad \Rightarrow \quad m(x) \leq m(y) \quad (x, y \in S).$$

We say that S is *bounded from below* resp. *above* if there exists an element 0 resp. 1 (necessarily unique) such that

$$0 \leq x \quad (x \in S) \quad \text{resp.} \quad x \leq 1 \quad (x \in S).$$

Finite lattices are bounded from below and above.

A map $m : S \rightarrow S$ is *additive* if

$$m(0) = 0 \quad \text{and} \quad m(x \vee y) = m(x) \vee m(y) \quad (x, y \in S).$$

Additive maps are monotone.

Monotone and additive duality

Let S be a partially ordered set (lattice), let $T := \{0, 1\}$, and let

$$\mathcal{F}_{\text{mon}}(S, T) := \{f \in \mathcal{F}(S, T) : f \text{ is monotone}\},$$

$$\mathcal{F}_{\text{add}}(S, T) := \{f \in \mathcal{F}(S, T) : f \text{ is additive}\}.$$

Let X be a Markov process with state space S and generator

$$Gf(x) = \sum_{m \in \mathcal{G}} r_m (f(m(x)) - f(x)).$$

If all maps $m \in \mathcal{G}$ are monotone (resp. additive), then

$\mathbb{X}_{s,t}$ is monotone (resp. additive) and hence

$f \circ \mathbb{X}_{s,t}$ is monotone (resp. additive) for all

$f \in \mathcal{F}_{\text{mon}}(S, T)$ (resp. $f \in \mathcal{F}_{\text{add}}(S, T)$).

Thus $\mathcal{F}_{\text{mon}}(S, T)$ (resp. $\mathcal{F}_{\text{add}}(S, T)$) is invariant.

Let S be a partially ordered set. A *dual* of S is a partially ordered set S' together with a bijection $S \ni x \mapsto x' \in S'$ such that

$$x \leq y \quad \text{if and only if} \quad x' \geq y'.$$

Example 1: For any partially ordered set S , we may take $S' := S$ but equipped with the reversed order, and $x \mapsto x'$ the identity map.

Example 2: If Λ is a set and $S \subset \mathcal{P}(\Lambda)$ is a set of subsets of Λ , equipped with the partial order of inclusion, then we may take for $x' := \Lambda \setminus x$ the complement of x and $S' := \{x' : x \in S\}$.

Additive duality

Let S be a finite lattice and let S' be its dual. Define $\psi : S \times S' \rightarrow T = \{0, 1\}$ by

$$\psi(x, y) = 1_{\{x \not\leq y'\}} = 1_{\{y \not\leq x'\}} \quad (x \in S, y \in S').$$

(Additive functions) For each $f \in \mathcal{F}_{\text{add}}(S, T)$, there exists a unique $y \in S'$ such that

$$f(x) = \psi(x, y) \quad (x \in S).$$

(Additive duality) A map $m : S \rightarrow S$ has a dual $m' : S' \rightarrow S'$ w.r.t. ψ if and only if m is additive. The dual map m' is unique and also an additive map.

Let $S = \{0, \dots, n\}$ be totally ordered and let $S' := S$ equipped with the reversed order.

A map $m : S \rightarrow S$ is additive iff m is monotone and $m(0) = 0$. Each such map has a dual $m' : S' \rightarrow S'$ that is monotone and satisfies $m'(n) = n$.

(Siegmund's dual) *Let X be a monotone Markov process in S such that 0 is a trap. Then X has a dual Y w.r.t. to the duality function $\psi(x, y) := 1_{\{x \leq y\}}$. The dual process is also monotone and has n as a trap. Moreover, the duality can be realized in a pathwise way.*

Additive particle systems

Let $S = \mathcal{P}(\Lambda)$ with Λ a finite set, and let $x \mapsto x' \in S' := \mathcal{P}(\Lambda)$ denote the complement map $x' := \Lambda \setminus x$.

Then $1_{\{x \not\subset y'\}} = 1_{\{x \cap y \neq \emptyset\}}$.

(Additive particle systems) *Let X be a Markov process in S whose generator can be represented in additive maps. Then X has a pathwise dual Y w.r.t. to the duality function $\psi(x, y) := 1_{\{x \cap y \neq \emptyset\}}$, and Y is also an additively representable Markov process.*

Examples: Voter model, contact process, exclusion process, systems of coalescing random walks.

Krone's duality

Steve Krone [AAP 1999] has studied a two-stage contact process, with state space of the form $S = \{0, 1, 2\}^\Lambda$.

He interprets $x(i) = 0, 1$, or 2 as an empty site, young, or adult organism, and defines maps

grow up	a_i	$\dots 1 \dots \mapsto \dots 2 \dots$
give birth	b_{ij}	$\dots 2 0 \dots \mapsto \dots 2 1 \dots$
young dies	c_i	$\dots 1 \dots \mapsto \dots 0 \dots$
death	d_i	$\dots 1 \dots \mapsto \dots 0 \dots$
or		$\dots 2 \dots \mapsto \dots 0 \dots$
grow younger	e_i	$\dots 2 \dots \mapsto \dots 1 \dots$

where in all cases not mentioned, the maps have no effect.

Krone's duality

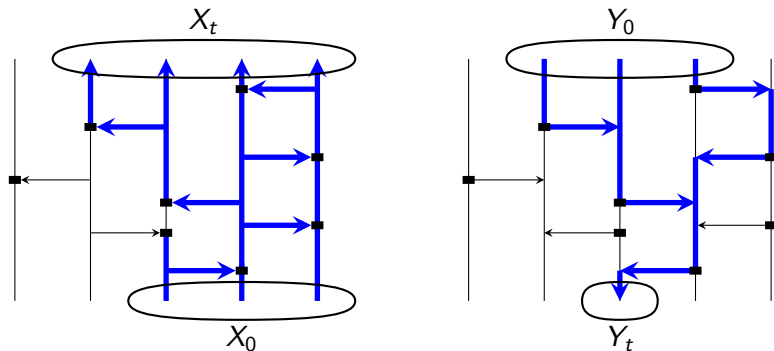
We set $S' := S$ and define $S \ni x \mapsto x' \in S'$ by $x'(i) := 2 - x(i)$. Then the duality function becomes

$$\psi(x, y) = 1_{\{x \not\leq y'\}} = 1_{\{\exists i \in \Lambda \text{ s.t. } x(i) + y(i) > 2\}}.$$

(Krone's dual) *The maps $a_i, b_{ij}, c_i, d_i, e_i$ are all additive and their duals are given by*

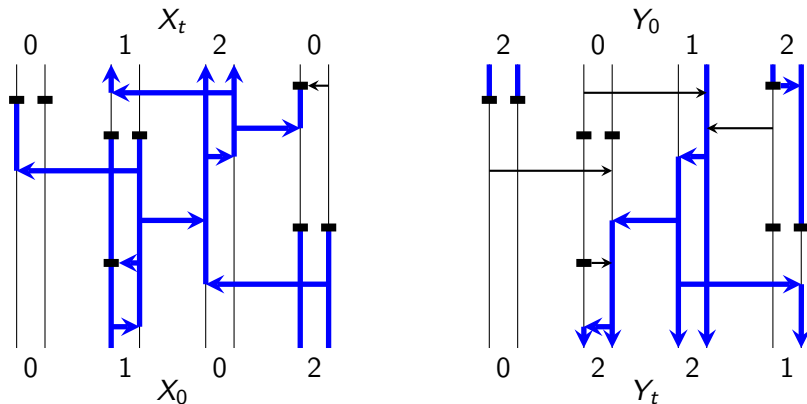
$$a'_i = a_i, \quad b'_{ij} = b_{ji}, \quad c'_i = e_i, \quad d'_i = d_i, \quad e'_i = c_i.$$

Percolation representations



Additive particle systems and their duals can be constructed in terms of open paths. In this example, X is a voter model and Y are coalescing random walks.

Percolation representations



We can also give a percolation representation of Krone's duality.

Percolation representations

By definition, a lattice S is *distributive* if

$$x \wedge (y \vee z) = (x \wedge y) \vee (x \wedge z) \quad (x, y, z \in S).$$

If Λ is a partially ordered set, then $S := \mathcal{P}_{\text{dec}}(\Lambda)$ with the order of set inclusion is a distributive lattice. *Birkhoff's representation theorem* says that every distributive lattice is of this form.

(Percolation representation) *An additive Markov process taking values in $\mathcal{P}_{\text{dec}}(\Lambda)$ has a percolation representation together with its dual, which takes values in $S' = \mathcal{P}_{\text{inc}}(\Lambda)$, with the duality function $\psi(x, y) = 1_{\{x \cap y \neq \emptyset\}}$.*

If Λ is equipped with the trivial order $x \not\leq y$ for all $x \neq y$, then $\mathcal{P}_{\text{dec}}(\Lambda) = \mathcal{P}(\Lambda) = \mathcal{P}_{\text{inc}}(\Lambda)$.

In Krone's example, $\Lambda = \{1, 2\}^\Delta$ with the product order.

Interacting particle systems

Let Λ be countable and let $S := \{0, 1\}^\Lambda$, equipped with the product topology. A map $m : S \rightarrow S$ is *local* if

- (i) m is continuous,
- (ii) $\{i \in \Lambda : \exists x \in S \text{ s.t. } m(x)(i) \neq x(i)\}$ is finite.

Note (i) is equivalent to:

- (i)' for each $i \in \Lambda$, the function $x \mapsto m(x)(i)$ depends on finitely many coordinates.

Assume all $m \in \mathcal{G}$ are local. Under suitable assumptions, the interacting particle system X with generator

$$Gf(x) = \sum_{m \in \mathcal{G}} r_m (f(m(x)) - f(x))$$

can be constructed from a stochastic flow $(\mathbb{X}_{s,u})_{s \leq u}$ based on a graphical representation ω .

Monotone systems duality

Notation:

$$\mathcal{C}(S, R) := \{f \in \mathcal{F}(S, R) : f \text{ is continuous}\},$$

$$\mathcal{L}(S, R) := \{f \in \mathcal{F}(S, R) : f \text{ is lower semi-continuous}\}.$$

Also: $\mathcal{C}_{\text{mon}}(S, R) := \mathcal{C}(S, R) \cap \mathcal{F}_{\text{mon}}(S, R)$ etc.

If all local maps $m \in \mathcal{G}$ are monotone, then a.s.

$$\mathbb{X}_{s,u} \in \mathcal{C}_{\text{mon}}(S, S) \quad (s \leq u).$$

Let $T := \{0, 1\}$. Then

$$f \in \mathcal{C}_{\text{mon}}(S, T) \quad \Rightarrow \quad f \circ \mathbb{X}_{s,u} \in \mathcal{C}_{\text{mon}}(S, T),$$

$$f \in \mathcal{L}_{\text{mon}}(S, T) \quad \Rightarrow \quad f \circ \mathbb{X}_{s,u} \in \mathcal{L}_{\text{mon}}(S, T).$$

Thus, $\mathcal{L}_{\text{mon}}(S, T)$ is preserved under $(\mathbb{F}_{s,u})_{s \leq u}$.

Monotone systems duality

Recall that $Y^\uparrow := \{z \in S : z \geq y \text{ for some } y \in Y\}$. One has

$$\mathcal{L}_{\text{mon}}(S, T) = \{1_Y : Y \in \mathcal{I}\}$$

with $\mathcal{I} := \{Y \in \mathcal{P}(S) : Y \text{ is open and } Y^\uparrow = Y\}$.

A convenient way to encode an open, increasing set is by writing down its minimal elements. A *minimal element* is an $y \in Y$ such that

$$z \in Y, z \leq y \text{ implies } z = y.$$

For each $Y \subset S$, let

$$Y^\circ := \{y \in Y : y \text{ is a minimal element of } Y\}.$$

It is easy to see that $(Y^\uparrow)^\uparrow = Y^\uparrow$ and $(Y^\circ)^\circ = Y^\circ$.

Monotone systems duality

$$\text{Let } S_{\text{fin}} := \{y \in S : |y| < \infty\} \quad \text{with} \quad |y| := \sum_{i \in \Lambda} y(i),$$

$$\mathcal{H} := \{Y \in \mathcal{P}(S_{\text{fin}}) : Y^\circ = Y\}.$$

(Encoding open increasing sets) *The map $Y \mapsto Y^\uparrow$ is a bijection from \mathcal{H} to \mathcal{I} , and $Y \mapsto Y^\circ$ is its inverse.*

Define $\psi : S \times \mathcal{H} \rightarrow T$ by

$$\psi(x, Y) := 1_{\{\exists y \in Y \text{ s.t. } x \geq y\}} \quad (x \in S, Y \in \mathcal{H}).$$

Then $\mathcal{H} \ni Y \mapsto \psi(\cdot, Y) \in \mathcal{L}_{\text{mon}}(S, T)$ is a bijection.

There exists a unique stochastic flow $(\mathbb{Y}_{s,u})_{s \leq u}$ on \mathcal{H} such that

$$\psi(\mathbb{X}_{s,u}^-(x), y) = \psi(x, \mathbb{Y}_{-u, -s}(Y)) \quad (x \in S, Y \in \mathcal{H}).$$

Monotone systems duality

Define constant configurations $\underline{0}(i) := 0$, $\underline{1}(i) := 1$ ($i \in \Lambda$).
By monotonicity, the process X has an *upper invariant law*

$$\mathbb{P}^{\underline{1}}[X_t \in \cdot] \xrightarrow[t \rightarrow \infty]{} \bar{\nu}.$$

By definition, the \mathcal{H} -valued process Y *survives* if

$$\mathbb{P}^{\{e_i\}}[Y_t \neq \emptyset \ \forall t \geq 0] > 0 \text{ for some } i \in \Lambda,$$

where $e_i(j) := 1_{\{i=j\}}$ ($i, j \in \Lambda$).

(Nontrivial upper invariant law) One has $\bar{\nu} \neq \delta_{\underline{0}}$ if and only if Y survives. The law $\bar{\nu}$ is uniquely characterized by

$$\mathbb{E}[\psi(\bar{X}, \{y\})] = \mathbb{P}^{\{y\}}[Y_t \neq \emptyset \ \forall t \geq 0].$$

where \bar{X} denotes a r.v. with law $\bar{\nu}$.

Proof

$$\begin{aligned}\mathbb{E}^{\underline{1}}[\psi(X_t, \{y\})] &= \mathbb{E}^{\{y\}}[\psi(\underline{1}, Y_t)] = \mathbb{E}^{\{y\}}[\exists y \in Y_t \text{ s.t. } \underline{1} \geq y] \\ &= \mathbb{E}^{\{y\}}[Y_t \neq \emptyset] \xrightarrow{t \rightarrow \infty} \mathbb{P}^{\{y\}}[Y_t \neq \emptyset \ \forall t \geq 0].\end{aligned}$$



Monotone systems duality

Equip the space \mathcal{H} with an order such that

$$Y \leq Z \quad \Leftrightarrow \quad \psi(x, Y) \leq \psi(x, Z) \quad \forall x \in S,$$

and with a topology such that

$$Y_n \rightarrow Y \quad \Leftrightarrow \quad \psi(x, Y_n) \rightarrow \psi(x, Y) \quad \forall x \in S_{\text{fin}}.$$

The minimal and maximal elements of \mathcal{H} are \emptyset and $\{\underline{0}\}$ since

$$\psi(x, \emptyset) = 0 \quad \psi(x, \{\underline{0}\}) = 1 \quad (x \in S).$$

The *second largest* element of \mathcal{H} is $Y_{\text{sec}} := \{e_i : i \in \Lambda\}$, since

$$\psi(x, Y_{\text{sec}}) = 1_{\{x \neq \underline{0}\}} \quad (x \in S).$$

(Recall $e_i(j) := 1_{\{i=j\}}$ ($i, j \in \Lambda$).)

Monotone systems duality

We say that X *survives* if

$$\mathbb{P}^{e_i}[X_t \neq \underline{0} \ \forall t \geq 0] > 0 \text{ for some } i \in \Lambda.$$

If $\underline{0}$ is a trap for $(X_t)_{t \geq 0}$, then $Y_0 \neq \{\underline{0}\}$ implies $Y_t \neq \{\underline{0}\}$ ($t \geq 0$).

Work in progress We believe that

$$\mathbb{P}^{Y_{\text{sec}}}[Y_t \in \cdot] \xrightarrow[t \rightarrow \infty]{} \bar{\mu}.$$

We call $\bar{\mu}$ the *upper invariant law* of $(Y_t)_{t \geq 0}$.

We believe that $\bar{\mu} \neq \delta_{\emptyset}$ if and only if X survives.

The law $\bar{\mu}$ is uniquely characterized by

$$\mathbb{E}\left[\prod_{k=1}^n \psi(x_k, \bar{Y})\right] = \mathbb{P}[\mathbb{X}_{0,t}(x_k) \neq \underline{0} \ \forall t \geq 0, \ k = 1, \dots, n].$$

where \bar{Y} denotes a r.v. with law $\bar{\mu}$ and $x_1, \dots, x_n \in S_{\text{fin}}$

Some simulations

i^\uparrow	
i	i^\rightarrow

For each $i = (i_1, i_2) \in \mathbb{Z}^2$, let
 $i^\rightarrow := (i_1 + 1, i_2)$ and $i^\uparrow := (i_1, i_2 + 1)$.

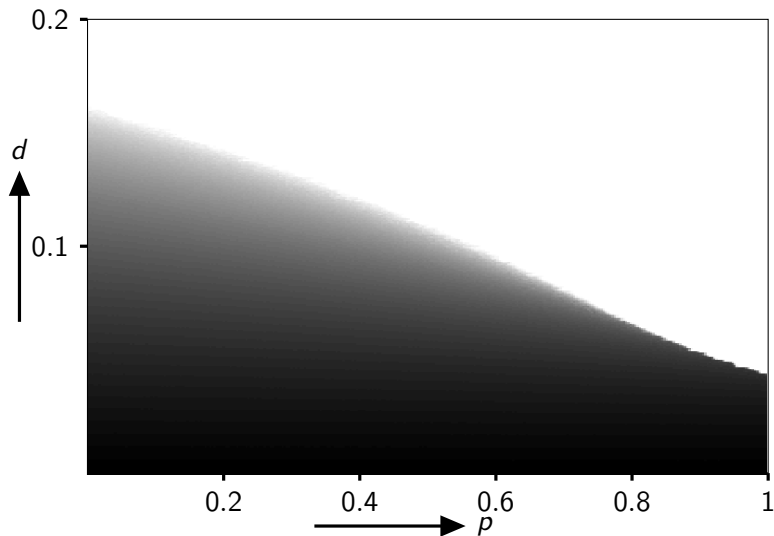
Let $p, d \in [0, 1]$ and let $X = (X_t)_{t \in \mathbb{N}}$ be a Markov chain with values in $\{0, 1\}^{\mathbb{Z}^2}$ such that independently for each i and t ,

$$\begin{aligned} X_{t+1}(i) &= X_t(i) \vee (X_t(i^\rightarrow) \wedge X_t(i^\uparrow)) && \text{w. prob. } p(1-d), \\ X_{t+1}(i) &= X_t(i) \vee X_t(i^\rightarrow) && \text{w. prob. } \frac{1}{2}(1-p)(1-d), \\ X_{t+1}(i) &= X_t(i) \vee X_t(i^\uparrow) && \text{w. prob. } \frac{1}{2}(1-p)(1-d), \\ X_{t+1}(i) &= 0 && \text{w. prob. } d. \end{aligned}$$

For $p = 0$ this model is additive.

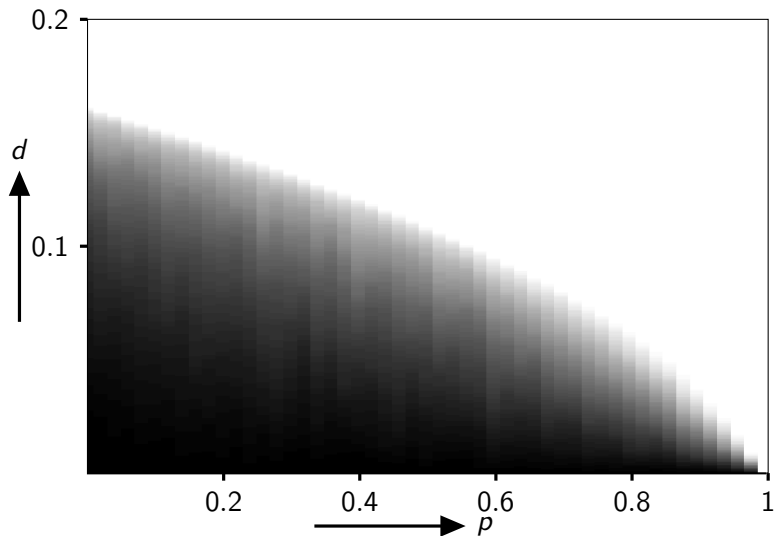
For $p = 1$, it does not survive for any $d > 0$.

Some simulations



Density of the upper invariant law.

Some simulations



Survival probability started from a single one.