Permanental processes

Michael B. Marcus Joint work with Jay Rosen

Department of Mathematics, CUNY

BIRS conference: High Dimensional Probability October 9–14, 2011. Let *T* be an index set and $\{G(x), x \in T\}$ be a mean zero Gaussian process with covariance $u(x, y), x, y \in T$. The process G^2 can be defined by the Laplace transform of its finite joint distributions

$$E\left(\exp\left(-\frac{1}{2}\sum_{i=1}^{n}\alpha_{i}G^{2}(x_{i})\right)\right)=\frac{1}{|I+\alpha U|^{1/2}}$$
(1)

for all x_1, \ldots, x_n in T, where I is the $n \times n$ identity matrix, α is the diagonal matrix with $(\alpha_{i,i} = \alpha_i)$, $\alpha_i \in R_+$ and $U = \{u(x_i, x_j)\}$ is an $n \times n$ matrix, that is symmetric and positive definite. (It is the covariance of $(G(x_1), \ldots, G(x_n))$.)

In 1997, D. Vere-Jones introduced the permanental process $\theta := \{\theta_x, x \in T\}$, which is a real valued positive stochastic process with finite joint distributions that satisfy

$$E\left(\exp\left(-\frac{1}{2}\sum_{i=1}^{n}\alpha_{i}\theta_{x_{i}}\right)\right) = \frac{1}{|I+\alpha\Gamma|^{\beta}},$$
(2)

where $\Gamma = {\Gamma(x_i, x_j)}_{i,j=1}^n$ is an $n \times n$ matrix and $\beta > 0$. (We refer to this as a β -permanental process.)

The generalization here is that Γ need not be symmetric.

For n = 2, and $\beta = 1/2$, (2) takes the form

$$\begin{split} & E\left(\exp\left(-\frac{1}{2}\left(\alpha_{1}\theta_{x}+\alpha_{2}\theta_{y}\right)\right)\right) \\ &= \frac{1}{|I+\alpha\Gamma|^{1/2}} = \left(1+\alpha_{1}\Gamma(x,x)+\alpha_{2}\Gamma(y,y)\right) \\ &+ \alpha_{1}\alpha_{2}\left(\Gamma(x,x)\Gamma(y,y)-\Gamma(x,y)\Gamma(y,x)\right)\right)^{-1/2}. \end{split}$$

For permanental processes

$$\Gamma(x,x) \ge 0, \qquad \Gamma(x,y)\Gamma(y,x) \ge 0.$$

and

$$\Gamma(x,x)\Gamma(y,y) - \Gamma(x,y)\Gamma(y,x) \ge 0.$$

Therefore, the matrix

$$\begin{bmatrix} \Gamma(x,x) & (\Gamma(x,y)\Gamma(y,x))^{1/2} \\ (\Gamma(x,y)\Gamma(y,x))^{1/2} & \Gamma(y,y) \end{bmatrix}$$

is positive definite, so that we can construct a mean zero Gaussian vector $\{G(x), G(y)\}$ with covariance matrix

$$E\left(G(x)G(y)\right) = \left(\Gamma(x,y)\Gamma(y,x)\right)^{1/2}.$$
(3)

WE MAY HAVE $\Gamma(x, y) \neq \Gamma(y, x)$

Set

$$d(x, y) = (E(G(x) - G(y))^2)^{1/2}$$

= $(\Gamma(x, x) + \Gamma(y, y) - 2(\Gamma(x, y)\Gamma(y, x))^{1/2})^{1/2}$

Lemma

Suppose that $\theta := \{\theta_x, x \in T\}$ is a 1/2-permanental process with kernel Γ . Then for any pair x, y,

$$\{\theta_x, \theta_y\} \stackrel{\text{law}}{=} \{G^2(x), G^2(y)\}$$
(4)

where $\{G(x), G(y)\}$ is a mean zero Gaussian random variable with covariance matrix given by (3).

The function d(x, y) is a metric, or pseudo-metric, on *T*, (although $(\Gamma(x, y)\Gamma(y, x))^{1/2}$ may not be positive definite.)

Let (T, d) be a separable metric or pseudometric space. Let $B_d(t, u)$ denote the closed ball in (T, d) with radius u and center t. For any probability measure μ on (T, d) we define

$$J_{\mathcal{T},d,\mu}(a) = \sup_{t\in\mathcal{T}}\int_0^a \left(\lograc{1}{\mu(B_d(t,u))}
ight)^{1/2}\,du.$$

A sufficient condition for continuity

Theorem

Let $\theta = \{\theta_x : x \in T\}$ be a 1/2-permanental process, with kernel Γ satisfying $\sup_{x \in T} \Gamma(x, x) < \infty$. Let D denote the d diameter of T and assume that T is separable for d, and that there exists a probability measure μ on $\mathcal{B}(T, d)$ such that

 $J_d(D) < \infty$.

Then there exists a version $\theta' = \{\theta'_x, x \in T\}$ of θ which is bounded almost surely.

Theorem (continued)

$$\lim_{\delta\to 0} J_d(\delta) = 0$$

there exists a version $\theta' = \{\theta'_x, x \in T\}$ of θ such that

$$\lim_{\delta \to 0} \sup_{\substack{s,t \in T \\ d(s,t) \le \delta}} |\theta_s'(\omega) - \theta_t'(\omega)| = 0, \qquad a.s.$$

If (3) holds and

$$\lim_{\delta \to 0} \frac{J_d(\delta)}{\delta} = \infty,$$

then

lf

Theorem (continued)

$$\lim_{\delta\to 0} \sup_{\substack{s,t\in T\\ d(s,t)\leq \delta}} \frac{|\theta_s'-\theta_t'|}{J_d(d(s,t)/2)} \leq 30 \left(\sup_{x\in T} \theta_x'\right)^{1/2} \quad a.s.$$

The proof is immediate. It follows from Lemma 1 that

$$\hat{d}(x,y) := \|\theta_x^{1/2} - \theta_y^{1/2}\|_{\psi_2} = \| |G_x| - |G_y| \|_{\psi_2}$$

$$\leq \| G_x - G_y \|_{\psi_2} = d(x,y).$$
(5)

Also $|\theta_s - \theta_t| \le |\theta_s^{1/2} - \theta_t^{1/2}| (\sup_{x \in T} \theta_x')^{1/2}$

Verre-Jones shows that a sufficient condition for (2) to hold is that all the real non-zero eigenvalues of Γ are positive and that $r\Gamma(I + r\Gamma)^{-1}$ has only non-negative entries for all r > 0. Eisenbaum and Kaspi note that this is the case when $\Gamma(x, y)$, $x, y \in T$, is the potential density of a transient Markov process on *T*. This enables them to find a Dynkin type isomorphism for the local times of Markov processes that are not necessarily symmetric, in which the role of G^2 is taken by the permanental process θ .

Permanental processes associated with Lévy processes

Certain kernels of permanental processes are associated with Lévy processes. Let $X = \{X_t, t \in R_+\}$ be a Lévy process with characteristic function

$$Ee^{i\lambda X_t} = e^{-\psi(\lambda)t}.$$
 (6)

Assume that X has local times $\{L_t^x, (x, t) \in \mathbb{R} \times \mathbb{R}_+\}$. Set

$$u_{T_0}(x,y) = E^x \left(L^y_{T_0} \right), \qquad (7)$$

where T_0 is the first hitting time of X at zero.

The function $u_{T_0}(x, y)$ is the zero potential of the transient Markov process $\widetilde{X} = {\widetilde{X}_t}$, which is X killed at the first time it hits zero, and thus is also the kernel of a permanental process.

Lemma

$$u_{T_0}(x,y) = R(x,y) + H(x,y)$$

and

$$u_{T_0}(y,x) = R(x,y) - H(x,y)$$

where

Lemma (continued)

$$\begin{aligned} R(x,y) &= R(y,x) = \frac{1}{\pi} \int_0^\infty \\ \frac{(1 - \cos \lambda x - \cos \lambda y + \cos \lambda (x - y)) \mathcal{R} e \psi(\lambda)}{|\psi(\lambda)|^2} \, d\lambda \end{aligned}$$

and

$$H(x,y) = -H(y,x) = \frac{1}{\pi} \int_0^\infty \frac{(\sin \lambda x - \sin \lambda y - \sin \lambda (x - y))\mathcal{I}m\psi(\lambda)}{|\psi(\lambda)|^2} d\lambda.$$

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Following Le Jan, (for symmetric Markov processes), we can identify the peramnental process as an explicit process called a loop soup local time.

Let *S* a be locally compact set with a countable base. Let $Y = (\Omega, Y_t, P^x, \mathcal{F}_t)$ be a recurrent Markov process with state space *S*, and jointly measurable transition densities $p_t(x, y)$ with respect to some σ -finite measure *m* on *S*. We assume that the 1-potential densities

$$u^{1}(x,y) = \int_{0}^{\infty} e^{-t} p_{t}(x,y) dt \qquad (8)$$

are continuous. We do not require that $u^1(x, y)$ is symmetric.

For all $0 < t < \infty$, and $x \in S$, there exists a finite measure $P_t^{x,x}$ on \mathcal{F}_{t^-} , of total mass $p_t(x, x)$, such that

$$P_{t}^{x,x}(F) = P^{x}(F p_{t-s}(Y_{s}, x)), \qquad (9)$$

for all $F \in \mathcal{F}_s$ with s < t.

For $\Delta \notin S$, let Ω_{Δ} denote the set of right continuous paths ω in $S \cup \Delta$ with $\omega_t = \Delta$ for all $t \ge \zeta$. We set $Y_t(\omega) = \omega_t$ and

$$\zeta = \inf\{t > 0 \mid, Y_t = \Delta\}.$$
(10)

The killing time ζ is determined by an operator $k_t \omega(s) = \omega(s)$ if s < t and $k_t \omega(s) = \Delta$ if $s \ge t$.

We define a σ -finite measure μ on $(\Omega_{\Delta}, \mathcal{F})$ by

$$\mu(\boldsymbol{A}) = \int_0^\infty \frac{\boldsymbol{e}^{-t}}{t} \int \boldsymbol{P}_t^{\boldsymbol{x},\boldsymbol{x}} \left(\boldsymbol{k}_t^{-1}(\boldsymbol{A}) \right) \, \boldsymbol{dm}(\boldsymbol{x}) \, \boldsymbol{dt}, \quad \boldsymbol{A} \in \mathcal{F}.$$
(11)

We refer to μ as the loop measure associated with the Markov process *Y*. Under certain regularity assumptions on *Y*, μ is supported on

$$\mathcal{L} = \{ Y : Y_{\zeta^{-}} = Y_0 \}.$$
 (12)

We call μ the loop measure for the Markov process *Y*.

Let \mathcal{L}_{α} be a Poisson point process on Ω_{Δ} with intensity measure $\alpha\mu$. Each realization of the random variable \mathcal{L}_{α} is countable subset of Ω_{Δ} . I.e. let

$$N(A) := \# \{ \mathcal{L}_{\alpha} \cap A \}, \quad A \subseteq \Omega_{\Delta}.$$
(13)

Then for any disjoint measurable subsets A_1, \ldots, A_n of Ω_{Δ} , the random variables $N(A_1), \ldots, N(A_n)$, are independent, and N(A) is a Poisson random variable with parameter $\alpha \mu(A)$, i.e.

$$P(N(A) = k) = \frac{(\alpha \mu(A))^k}{k!} e^{-\alpha \mu(A)}.$$
 (14)

The Poisson point process \mathcal{L}_{α} is called the 'loop soup' of the Markov process *Y*.

We define the 'loop soup local time', \hat{L}^x , of Y, by

$$\widehat{L}_{\alpha}^{x} = \sum_{\omega \in \mathcal{L}_{\alpha}} \ell^{x}(\omega), \tag{15}$$

where $\ell^{x}(\omega)$ is the local time of the path $\omega \in \Omega_{\Delta}$.

Theorem

Let $\{\widehat{L}_{\alpha}^{x}, x \in S\}$ be the loop soup local time of Y. Then $\{2\widehat{L}_{\alpha}^{x}, x \in S\}$, is an α -permanental process with kernel $u^{1}(x, y)$.