Potential Theory, Self-Similarity & Statistical Mechanics

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Outline

- 1. Review of some Stochastic Potential Theory
 - Example: Random Interlacements
- 2. Self-Similar Markov Processes
 - Example: Hausdorff Dimensions of "Self-Similar Sets"

Part 1: Review of some Stochastic Potential Theory

Principal Objects

• locally compact statespace (E, \mathcal{E}) ,

- Markov process (X_t) with transition semigroup (P_t)
- excessive measure on (E, \mathcal{E}) for (P_t) , i.e. σ -finite and

$$P^{\eta}(X_t \in \cdot) =: \eta P_t(\cdot) \leq \eta(\cdot)$$

Special cases:

- invariant measure: $\eta P_t = \eta$
- purely excessive: $\eta P_t(A) \downarrow 0$ for $t \to \infty$

Examples:

• $\eta(dx) = dx$ for Brownian motion

• potential measures $\eta_{\nu}(A) = \mathbb{E}^{\nu} [\int_{0}^{\infty} 1_{A}(X_{t}) dt]$ are purely excessive

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Research Directions

- Structure of excessive measures Example: Riesz decomposition of η into invariant and purely excessive parts: $\eta = \eta_I + \eta_P$.
- Relation of excessive measures and path-behavior (entrance boundaries)
- Martin boundary

...

Trajectories with Birth & Death

Add to E a cemetary state ∂ . Set

 $D = \left\{ Y : \mathbb{R} \to E \cup \{\partial\} \ \mathsf{RCLL} \mid Y_t \in E \ \mathsf{for} \ t \in (\alpha(w), \beta(w)) \right\}$

and call

- $\alpha(Y)$ time of birth of Y,
- $\beta(Y)$ time of death of Y,
- $\zeta(Y) = \beta(Y) \alpha(Y)$ life-time of Y.

Processes with Birth & Death - Kuznetsov Measures

Measure \mathcal{Q}_η on (D,\mathcal{D}) is called Kuznetsov measure for

- Markov process (*P*_t)
- excessive measure η

if \mathcal{Q}_{η} is a Markov process with birth & death and marginals η .

That is, for $-\infty < t_1 < \cdots < t_n < +\infty$

$$\mathcal{Q}_{\eta}(\alpha(Y) < t_1, Y_{t_1} \in dx_1, \cdots, Y_{t_n} \in dx_n, t_n < \beta(Y))$$

= $\eta(dx_1)P_{t_2-t_1}(x_1, dx_2) \cdots P_{t_n-t_{n-1}}(x_{n-1}, dx_n).$

Theorem (Kuznetsov): Existence and Uniqueness.

Processes with Birth & Death - Quasi Processes

Measure \mathcal{P}_{η} on (D, \mathcal{D}) - modulus time translation - is called quasi process for

- Markov process (*P*_t)
- excessive measure η

if

(a) for all shift-invariant stopping times ${\boldsymbol S}$

 $(Y_{S+t})_{t\geq 0}$

is Markov with transitions (P_t) under $\mathcal{P}_{|_{S \in \mathbb{R}}}$.

(b) $\eta(A) = \mathcal{P}_{\eta}(\int_{\alpha}^{\beta} 1_{A}(Y_{s}) ds)$

Theorem (Weil, Hunt): Existence and Uniqueness.

Palm Measure Relation of \mathcal{Q}_{η} and \mathcal{P}_{η}

Theorem (Fitzsimmons): Fix any finite shift-invariant time S, then

$$\mathcal{P}_\eta(A) = \mathcal{Q}_\eta(A, S \in [0, 1])$$

for all shift-invariant events A.

Riesz Decomposition - Fitzsimmons/Maisonneuve

<u>Theorem</u>: Let η excessive and define

$$\mathcal{P}_{\mathcal{P}}(\cdot) := \mathcal{P}_{\eta}(\cdot, \alpha > -\infty), \qquad \mathcal{P}_{\mathcal{I}}(\cdot) := \mathcal{P}_{\eta}(\cdot, \alpha = -\infty).$$

Then $\eta = \eta_I + \eta_P$ with

$$\eta_P(A) = \mathcal{P}_P\Big(\int_{\alpha}^{\beta} 1_A(Y_s) ds\Big) \qquad \eta_I(A) = \mathcal{P}_I\Big(\int_{-\infty}^{\beta} 1_A(Y_s) ds\Big).$$

Remark: Analogous theorem for Kuznetsov measures.

Riesz and a "Way of starting stochastic processes"

starting process $(X_t) \iff$ purely excessive measure " \Rightarrow " occupation measure " \Leftarrow " take quasi process and shift to $\alpha = 0$ (finite measure?) "take" is abstract, often means "take Kuznetsov, then Palm measure" starting process (X_t) at time $-\infty \iff$ invariant measure Problem: Only know processes forwards in time, not backwards. Example: Lévy processes.

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Kuznetsov Measures and Duality

Definition: Two Markov processes (X_t) and (\hat{X}_t) are in duality with respect to a measure *m* if

$$P_t(x, dy)m(dx) = \hat{P}_t(y, dx)m(dy).$$

Examples:

- Brownian motion is self-dual with m(dx) = dx
- Lévy process (ξ_t) is dual to $(\hat{\xi}_t) = (-\xi_t)$ with m(dx) = dx

<u>Mitro's Theorem</u>: If *m* is invariant for (P_t) and (\hat{P}_t) is a dual process, then Q_m is constructed as follows:

- sample *m* at time 0
- run (P_t) in positive time direction
- run (\hat{P}_t) in negative time direction

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Example: Random Interlacements

Definition: Brownian interlacement is the (infinite) measure ν on (D, D)- modulus time translation - so that restricted to trajectories hitting balls $B \subset \mathbb{R}^n$, $n \ge 3$,

- first hitting distribution at B is harmonic measure
- forwards in time paths are Brownian motion
- backwards in time paths are conditioned Brownian motion

Brownian random interlacement is Poisson random measure (loop soup) with intensity $\boldsymbol{\nu}.$

Theorem (D., Dereich '14, Rosen '14):

Interlacement is quasi process of BM with invariant measure m(dx) = dx.

Corollary:

Interlacement is two-sided BM started in Lebesgue measure restricted to be closest to the origin at time $S \in [0, 1]$ modulus time translations.

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Part 2: Self-Similar Markov Processes

A real-valued Markov process is called self-similar if it scales in time and space:

the law of $(cX_{c^{-1/\gamma}t})_{t\geq 0}$ under \mathbb{P}_z is \mathbb{P}_{cz}

 γ is called the index of self-similarity.

Goals

(1) Understand the general form of a self-similar process.

(2) Deduce non-trivial consequences.

Seeking magic arguments of type

- know self-similarity for a model
- know "something additional" for the model

to deduce from general theory all kind of things.

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For simplicity now restrict ourselves to positive processes.

This is <u>NOT needed</u>, but enough for application in the end!

Lamperti's Representation

Behavior up to hitting zero governed by Lévy processes:

Theorem (Lamperti '72)

If (X_t) is positive self-similar there is a Lévy process (ξ_t) s.t. for z > 0

$$X_t^{(z)} := z \exp\left(\xi_{A(tz^{-\gamma})}\right), \quad t \leq T_0,$$

where

$$A(t) := \left(\int_0^t \exp\left(\gamma\xi_s\right) ds\right)^{-1}.$$

Lamperti's Dichotomy

 $\textcircled{0} \xi \text{ drifts to } -\infty \text{ if and only if}$

$$\mathbb{P}_z(T_0 < \infty) = 1$$

 \rightarrow "recurrent case"

2 ξ does not drift to $-\infty$ if and only if

$$\mathbb{P}_z(T_0=\infty)=1$$

 \rightarrow "transient case"

Example

Squared-Bessel processes of dimension $\delta \in \mathbb{R}$

$$dX_t = 2\sqrt{X_t}dB_t + \delta dt, \quad t \leq T_0,$$

self-similar of index 1 with corresponding Lévy process

$$\xi_t = 2B_t + (\delta - 2)t.$$

Lamperti's dichotomy:

- $\delta < 2 \Rightarrow$ squared-Bessel processes hit zero
- $\delta \ge 2 \Rightarrow$ squared-Bessel processes do not hit zero

- (R) Recurrent case: describe (if exists) the positive self-similar process after T_0 .
- (T) Transient case: construct (if $\lim_{z\to 0} \mathbb{P}_z = \mathbb{P}_0$ exists) the limit \mathbb{P}_0 !

Recurrent case via potential theory by Rivero '06, Fitzsimmons '05. Lamperti SDE approach in Barczy/D. '13.

Transient case gradually by Bertoin, Caballero, Chaumont, Kyprianou, Pardo, Rivero, Savov, Yor ('02-13')

Extensions for real processes in D. '13, Dereich, D., Kyprianou '14+

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Definition: We say ξ has stationary overshoots if

$$O := \lim_{x \to \infty} (\xi_{T_x} - x)$$
 exists non-trivially.

Remark: If $\xi \uparrow +\infty$, equivalent to $\mathbb{E}[|\xi_1|] < \infty$.

Theorem:

 $\lim_{z\to 0} \mathbb{P}_z =: \mathbb{P}_0$ exists if and only if ξ has stationary overshoots.

Proof (Necessity): For a < b

$$\lim_{z \to 0} \mathbb{P}_{z}(X_{T_{a}} \leq b)$$

$$= \lim_{z \to 0} P(z \exp(\xi_{T_{\log(a/z)}}) \leq b)$$

$$= \lim_{z \to 0} P(\xi_{T_{\log(a/z)}} \leq \log(b/z))$$

$$= \lim_{z \to 0} P(\xi_{T_{\log(a/z)}} - \log(a/z) \leq \log(b/a))$$
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Proof (Sufficiency): Needs a construction of \mathbb{P}_0 , harder.

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Idea: Take "the good" purely excessive measure η_0 and set

$$\mathbb{P}_0 = rac{\mathcal{P}_{\eta_0}}{P_{\eta_0}(\mathcal{D})}$$

shifted to $\alpha = 0$. Need $\mathcal{P}_{\eta_0}(\mathcal{D}) < \infty$ and left-limits 0!

Question: Which excessive measure η_0 ?

Examples: $\eta_{\beta}(dx) = x^{1/\gamma - 1 + \beta} dx$ are purely excessive for many $\beta \ge 0$.

Answer: $\eta_0(dx) = x^{1/\gamma-1}dx$ is the good one.

Problem: No idea how to prove this directly! No duality \rightarrow left limits?

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A Theorem of Kaspi

Setting:

- Markov processes with $\tilde{X}_t = X_{A_t^{-1}}$ with $A_t = \int_0^t f(X_s) \, ds$.
- Kuznetsov measure Q_m for X with invariant measure m
- $B_t := \int_{-\infty}^t f(Y_s) \, ds < \infty$ under \mathcal{Q}_m .

Theorem (Kaspi): There is a purely excessive measure $\tilde{\eta}$ for \tilde{X} so that the quasi process $\tilde{\mathcal{P}}_{\tilde{\eta}}$ shifted to $\alpha = 0$ is obtained from $\mathcal{Q}_m(\cdot, B_0^{-1} \in [0, 1])$ as $\tilde{Y}_t = Y_{B_t^{-1}}$.

Important: Know left limits for $Q_m \Rightarrow$ know left limits for $\tilde{\mathcal{P}\eta}$.

Construction of \mathbb{P}_0 (for $\xi \uparrow +\infty$):

- Set m(dx) = dx for ξ and take Kuznetsov measure Q_m Mitro's two-sided construction implies left limits lim_{t↓-∞} Y_t = -∞.
- Define $B_t = \int_{-\infty}^t \exp(\gamma Y_s) ds$ and show $B_t < \infty$.
- Use Kaspi's theorem, left limits remain $-\infty$.
- Take exponential, left limits are 0.
- Normalize to probability law \mathbb{P}_0 .

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Convergence $\lim_{z\to 0} \mathbb{P}_z =: \mathbb{P}_0$:

An Application to Random Sets

Proposition: Suppose $M \subset [0,\infty)$ is a random set with

(a) *M* is the range of some increasing self-similar Markov process issued from 0.

(b) For some
$$\kappa \in$$
 (4,8) and all $a < b$

$$\mathbb{P}(M\cap [a,b]\neq \emptyset)=C\int_0^{\frac{b-a}{a}}\frac{1}{u^{2-8/\kappa}(1-u)^{4/\kappa}}\,du.$$

Then

$$dim_H(M) = 2 - \frac{8}{\kappa}$$

almost surely.

Example: If γ is $SLE(\kappa)$ and $M = \gamma \cap [0, \infty)$, then (b) holds. Schramm/Zhou '08 and Alberts/Sheffield '08 proved $\dim_H(M) = 2 - \frac{8}{\kappa}$.

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Remark: If (a) holds for SLE, then our approach gives much more than only Hausdorff dimension.

From self-similarity: Recall the construction of \mathbb{P}_0 . Only need to know Hausdorff dimension of $Range(\xi)$.

From subordinator theory:

$$dim_{H}(Range(\xi)) = \sup \Big\{ \gamma \leq 1 : \lim_{b \downarrow 0} b^{\gamma-1} \int_{0}^{b} \Pi(r, \infty) dr = +\infty \Big\}.$$

From fluctuation theory: Get Π from overshoot distributions as

$$C\int_0^b \Pi(r,\infty) \, dr = \lim_{x\uparrow+\infty} P(\xi_{T_x} - x \le b) \tag{2}$$

Lemma: $\int_0^b \Pi(r,\infty) dr \sim Cb^{8/\kappa-1}$ as $b \downarrow 0$. Proof:

$$\int_{0}^{\frac{b-a}{a}} \frac{1}{u^{2-8/\kappa}(1-u)^{4/\kappa}} du = \mathbb{P}(M \cap [a, b] \neq \emptyset)$$

= $\mathbb{P}_{0}(X_{T_{a}} \leq b)$
= $\lim_{z \downarrow 0} \mathbb{P}_{z}(X_{T_{a}} \leq b)$
 $\stackrel{\text{above}}{=} \lim_{z \to 0} P(\xi_{T_{\log(a/z)}} - \log(a/z) \leq \log(b/a))$
 $\stackrel{\text{above}}{=} C \int_{0}^{\log(b/a)} \Pi(r, \infty) dr$

This gives $\Pi(r,\infty)$. Only need a = 1 to get

$$C \int_0^b \Pi(r,\infty) \, dr = \int_0^{e^b-1} rac{1}{u^{2-8/\kappa}(1-u)^{4/\kappa}} \, du \sim C b^{8/\kappa-1}.$$

Lemma: $dim_H(Range(\xi)) = 2 - \frac{8}{\kappa}$ almost surely. Proof:

$$dim_{H}(M) \stackrel{\text{above}}{=} dim_{H}(Range(\xi))$$

$$= \sup \left\{ \gamma \leq 1 : \lim_{b \downarrow 0} b^{\gamma - 1} \int_{0}^{b} \Pi(r, \infty) dr = +\infty \right\}$$

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$$= 2 - \frac{8}{\kappa}.$$

Remark: The calculations above also show that

$$\Pi(r,\infty) = C(1-e^{-r})^{8/\kappa-2}e^{(1-4/\kappa)r}.$$

Hence, any result for subordinators that involves the Lévy measure only gives a result for *M*.

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