# INFINITESIMAL GENERATORS OF POLYNOMIAL PROCESSES

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# 1 Algebra for polynomial processes

### Quadratic harnesses













































#### 2 Quadratic harnesses







### Algebra of sequences of polynomials

Let Q be a linear space of sequences of polynomials in variable  $x \in \mathbb{R}$ . For  $\mathbb{P} = (p_0, p_1, ..., )$  and  $\mathbb{Q} = (q_0, q_1, ...)$  we define  $\mathbb{P} \mathbb{Q} =: \mathbb{R} = (r_0, r_1, ...) \in Q$  where

$$r_k = \sum_j [q_k]_j p_j, \qquad k = 0, 1, \dots$$

This product is associative. The identity is

$$\mathbb{E}=(1,x,x^2,\ldots).$$

If  $\deg(p_n) = n$  for all  $n \ge 0$  then  $\mathbb{P} = (p_0, p_1, p_2, \ldots)$  is invertible.

### ${\mathbb F}$ and ${\mathbb D}$

We will need special elements  $\mathbb{F}$ ,  $\mathbb{D} \in \mathcal{Q}$ :

 $\mathbb{F} = (x, x^2, x^3, x^4 \dots)$  and  $\mathbb{D} = (0, 1, x, x^2, \dots).$ 

We note that

 $\mathbb{DF} = \mathbb{E}$  and  $\mathbb{E} - \mathbb{FD} = (1, 0, 0, 0, ...)$ 

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### Isomorphism

Let  $\mathcal{P}$  denote the space of all polynomials on  $\mathbb{R}$ .

The algebra  $\operatorname{End}(\mathcal{P})$  is isomorphic to the algebra  $\mathcal{Q}$ . The isomorphism  $\Psi : \operatorname{End}(\mathcal{P}) \to \mathcal{Q}$  is defined by

$$\Psi(\mathbf{P}) = (\mathbf{P}(1), \mathbf{P}(x), \mathbf{P}(x^2), \ldots) := \mathbb{P} \in \mathcal{Q}, \qquad \forall \, \mathbf{P} \in \mathrm{End}(\mathcal{P})$$

For  $\mathbf{P}, \, \mathbf{Q} \in \operatorname{End}(\mathcal{P})$  and  $\mathbb{P} = \Psi(\mathbf{P}), \, \mathbb{Q} = \Psi(\mathbf{Q})$ 

$$\Psi(\mathbf{P} \circ \mathbf{Q}) = \mathbb{PQ}.$$

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# **Polynomial process**

Let  $\mathcal{P}_{\leq k}$  be a space of polynomials on  $\mathbb{R}$  with degree at most k, k = 0, 1, 2, ...

Let  $X = (X_t)_{t \ge 0}$  be a non-homogeneous Markov process with infinite state space  $S \subset \mathbb{R}$ . If

 $\mathrm{E}(f(X_t)|X_s) \in \mathcal{P}_{\leq k}$ 

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for any  $f \in \mathcal{P}_{\leq k}$ , k = 1, 2, ..., we call X a polynomial process.

# Family of operators on $\mathcal{P}$

Through

$$\mathbf{P}_{s,t} f(\mathbf{x}) := \mathrm{E}(f(X_t)|X_s = \mathbf{x}), \qquad f \in \mathcal{P},$$

such process can be identified with a family of linear operators  $\mathbf{P}_{s,t} : \mathcal{P} \to \mathcal{P}, 0 \le s \le t$ , satisfying

• 
$$\mathbf{P}_{s,t}(\mathcal{P}_{\leq k}) = \mathcal{P}_{\leq k}, k \geq 0;$$

• 
$$P_{s,t}(1) = 1;$$

• for  $0 \le s \le t \le u$ 

$$\mathbf{P}_{s,t} \circ \mathbf{P}_{t,u} = \mathbf{P}_{s,u}.$$

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# Family of elements of $\mathcal{Q}$

Through isomorphism  $\Psi$  : End( $\mathcal{P}$ )  $\rightarrow \mathcal{Q}$  the process X can be identified with a family of  $\mathbb{P}_{s,t} = (p_{s,t}^0, p_{s,t}^1, p_{s,t}^2, \ldots) \in \mathcal{Q}$ ,  $0 \le s \le t$  satisfying

•  $\mathbb{P}_{s,t}$  is invertible;

• 
$$\mathbb{P}_{s,t}(\mathbb{E} - \mathbb{FD}) = \mathbb{E} - \mathbb{FD};$$

• for 
$$0 \le s \le t \le u$$

$$\mathbb{P}_{s,t}\mathbb{P}_{t,u}=\mathbb{P}_{s,u}.$$
 (1)

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# Martingale polynomials

From the above conditions we conclude that

• 
$$\mathbb{P}_{t,t} = \mathbb{E}$$
  
• if  $\mathbb{M}_t = \mathbb{P}_{0,t}^{-1}$  then  
 $\mathbb{M}_s = \mathbb{P}_{s,t}\mathbb{M}_t.$  (2)

The identity (2) follows by multiplying the flow equation

$$\mathbb{P}_{\mathbf{0},s}\mathbb{P}_{s,t}=\mathbb{P}_{\mathbf{0},t}$$

by  $\mathbb{M}_s$  and  $\mathbb{M}_t$ :

 $\mathbb{M}_{s}\mathbb{P}_{0,s}\mathbb{P}_{s,t}\mathbb{M}_{t}=\mathbb{M}_{s}\mathbb{P}_{0,t}\mathbb{M}_{t}.$ 

# Martingale polynomials, cont.

Condition (2) says that if  $\mathbb{M}_t = (m_t^0, m_t^1, m_t^2, ...)$  then  $(m_t^k)_{t \ge 0}$ ,  $k \ge 0$ , are martingale polynomials for *X*, that is

$$\operatorname{E}(m_t^k(X_t)|X_s)=m_s^k(X_s), \qquad 0\leq s\leq t \quad k\geq 0.$$

Note that  $\deg(m_t^k) = k, k \ge 0$ .

Note also that

$$\mathbb{P}_{s,t} = \mathbb{M}_s \mathbb{M}_t^{-1}$$

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### Harness (Hammersley, 1967, Mansuy, Yor, 2005)

#### Let $X = (X_t)_{t \ge 0}$ be a real valued stochastic process,

$$\mathbf{E} X_t = \mathbf{0}, \qquad \mathbf{E} X_s X_t = s \qquad \forall \mathbf{0} \le s \le t.$$

Let  $(\mathcal{F}_{s,u})_{0 \le s < u}$  be a natural **past-future filtration** of X, i.e.

$$\mathcal{F}_{\boldsymbol{s},\boldsymbol{u}}=\sigma\{\boldsymbol{X}_{\alpha},\,\alpha\not\in(\boldsymbol{s},\boldsymbol{u})\}.$$

The process *X* is a **harness** if  $\forall 0 \le s < t < u$ 

$$\mathbb{E}(X_t | \mathcal{F}_{s,u}) = a_{tsu} X_s + b_{tsu} X_u = \frac{(u-t)X_s + (t-s)X_u}{u-s} = t\Delta_{s,u} + \widetilde{\Delta}_{s,u},$$

where

$$\Delta_{s,u} = \frac{X_u - X_s}{u - s} \quad \text{oraz} \quad \widetilde{\Delta}_{s,u} = \frac{\frac{1}{u} X_u - \frac{1}{s} X_s}{\frac{1}{u} - \frac{1}{s}}.$$

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# Quadratic harness (BMW, 2007)

If X is a harness and additionally

$$\mathbb{E}(X_t^2|\mathcal{F}_{s,u}) = A_{tsu}X_s^2 + B_{tsu}X_sX_u + C_{tsu}X_u^2 + D_{tsu}X_s + E_{tsu}X_u + F_{tsu}X_u + F_{tsu}X_u$$

then X is called a **quadratic harness**.

Then there exist numbers  $\theta$ ,  $\eta \in \mathbb{R}$ ,  $\tau$ ,  $\sigma \ge 0$ ,  $\gamma \le 1 + 2\sqrt{\tau\sigma}$ such that  $\operatorname{Var}(X_t | \mathcal{F}_{s,u}) = F_{tsu} K\left(\Delta_{s,u}, \widetilde{\Delta}_{s,u}\right)$ ,

where

$$K(x, y) := 1 + \theta x + \eta y + \tau x^2 + \sigma y^2 - (1 - \gamma) xy$$
  
and 
$$F_{tsu} = \frac{(u-t)(u-s)}{u(1+\sigma s)+\tau-\gamma s}.$$

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$$\operatorname{Var}(X_t|\mathcal{F}_{s,u})=F_{tsu}\,\mathcal{K}\left(\Delta_{s,u},\,\widetilde{\Delta}_{s,u}\right),$$

where

$$\begin{split} \mathcal{K}(x,y) &:= 1 + \theta x + \eta y + \tau x^2 + \sigma y^2 - (1-\gamma)xy\\ \text{and} \qquad \mathcal{F}_{tsu} &= \frac{(u-t)(u-s)}{u(1+\sigma s) + \tau - \gamma s}. \end{split}$$

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# **Typical properties**

- linear conditional means and quadratic conditional variances (as above, except the case  $\gamma = 1 + 2\sqrt{\tau\sigma}$ ) uniquely determine all moments, and these moments uniquely determine the process  $X \sim \text{QH}(\theta, \eta, \tau, \sigma; \gamma)$ ;
- X is a (non-homogeneous) Markov process;
- X has orthogonal martingale polynomials  $(m_n(\cdot, t))$  i.e. for  $t \ge 0$

$$\operatorname{E} m_n(X_t,t) m_k(X_t,t) = 0, \quad k \neq n$$

and for  $n \ge 0$ 

$$\mathbb{E}(m_n(X_t,t)|\mathcal{F}_s) = m_n(X_s,s) \quad s < t.$$

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# Examples of $X \sim QH(\theta, \eta, \tau, \sigma, \gamma)$

- QH(0, 0, 0, 0, 1) Wiener process,
- $QH(\theta, 0, 0, 0, 1)$  centered Poisson process,
- QH(θ, 0, τ, 0, 1) Lévy-Meixner processes (Schoutens, 2000),
- QH(0, 0, 0, 0, 0) free Brownian motion (Biane, 1998),
- QH(0, 0, 0, 0, q) q-Gaussian process (Bożejko, Kümmerer, Speicher, 1997),
- QH(θ, 0, τ, 0, 0) free Lévy-Meixner process (Anshelevich, 2003),
- QH(θ, 0, τ, 0, q) q-Lévy-Meixner processes (BW, 2005)
- QH(θ, η, 0, 0, q) bi-Poisson processes (Biane, 1996, BW, 2006 (q = 1), BW 2007 (q = 0), BMW, 2008 (q ∈ [-1, 1]),
- $QH(\theta, \eta, \tau, \sigma, -\tau\sigma)$  free quadratic harness (BMW, 2011).











# Definition

 $(\mathbb{P}_{s,t}, 0 \le s \le t)$  is a quadratic harness  $QH(\theta, \eta, \tau, \sigma\gamma)$  flow if (martingale)  $\mathbb{P}_{s,t}(\mathbb{FD} - \mathbb{F}^2 \mathbb{D}^2) = \mathbb{FD} - \mathbb{F}^2 \mathbb{D}^2$ ,

2 (harness) There exists  $\mathbb{X} \in \mathcal{Q}$  such that

$$\mathbb{P}_{0,t}\mathbb{F} = (\mathbb{F} + t\mathbb{X})\mathbb{P}_{0,t},$$

(quadratic harness) The above X satisfies

$$\mathbb{XF} - \gamma \mathbb{FX} = \mathbb{E} + \theta \mathbb{X} + \eta \mathbb{F} + \tau \mathbb{X}^2 + \sigma \mathbb{F}^2.$$

### Martingale and harness

Ad.1. Note that  $\mathbb{FD} - \mathbb{F}^2 \mathbb{D}^2 = (0, x, 0, 0, ...)$ . That is  $\mathbf{P}_{s,t}(x) = x$ , meaning that  $\mathbb{E}(X_t | \mathcal{F}_s) = X_s$ .

Ad.2. For a martingale polynomial  $m_n(\cdot, t)$  the harness property gives

$$\mathbb{E}(X_t m_n(X_u, u) | X_s) = \frac{u-t}{u-s} X_s m_n(X_s, s) + \frac{t-s}{u-s} \mathbb{E}(X_u m_n(X_u, u) | X_s).$$

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## Ad 2, cont.

#### Equivalently,

$$\mathbb{P}_{s,t}\mathbb{F}\mathbb{M}_t = \frac{u-t}{u-s}\mathbb{F}\mathbb{M}_s + \frac{t-s}{u-s}\mathbb{P}_{s,u}\mathbb{F}\mathbb{M}_u.$$
(3)

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Define  $\mathbb{J}_t$  by  $\mathbb{FM}_t = \mathbb{M}_t \mathbb{J}_t$ . Then from (3) we get

$$(u-s)\mathbb{J}_t = (u-t)\mathbb{J}_s + (t-s)\mathbb{J}_u$$

and thus

$$\mathbb{J}_t = \mathbb{Y} + t\mathbb{X}$$
 for some  $\mathbb{X}, \mathbb{Y} \in \mathcal{Q}$ .

Consequently,

$$\mathbb{FM}_t = \mathbb{M}_t(\mathbb{Y} + t\mathbb{X}).$$

Since  $\mathbb{M}_0 = \mathbb{E}$  we get that  $\mathbb{Y} = \mathbb{F}$ .

#### Generator

The left infinitesimal generator  $\mathbb{A}_t^-$  exists since  $\mathbb{M}_t$  is element-wise differentiable in *t* 

$$\mathbb{A}_t^- \mathbb{M}_t = \lim_{h \to 0^+} \frac{1}{h} \left( \mathbb{P}_{t-h,t} - \mathbb{E} \right) \mathbb{M}_t = \lim_{h \to 0^+} \frac{1}{h} (\mathbb{M}_{t-h} - \mathbb{M}_t) = -\frac{\partial}{\partial t} \mathbb{M}_t.$$

The right infinitesimal generator exists since  $\mathbb{P}_{s,t}$  is continuous in *t* 

 $\mathbb{A}_{t}^{+}\mathbb{M}_{t} = \lim_{h \to 0^{+}} \frac{1}{h} (\mathbb{P}_{t,t+h}\mathbb{M}_{t} - \mathbb{M}_{t}) = \lim_{h \to 0^{+}} \mathbb{P}_{t,t+h} (\mathbb{M}_{t} - \mathbb{M}_{t+h})$  $= \lim_{h \to 0^{+}} \mathbb{P}_{t,t+h} \lim_{h \to 0^{+}} \frac{1}{h} (\mathbb{M}_{t} - \mathbb{M}_{t+h}) = -\frac{\partial}{\partial t} \mathbb{M}_{t}$ Consequently,

$$\mathbb{A}_t^- = \mathbb{A}_t^+ := \mathbb{A}_t \in \mathcal{Q}.$$

## Uniqueness

We note that for  $\mathbb{A}_t = (a_t^0, a_t^1, a_t^2, \ldots)$ 

**Proposition.** A polynomial process  $\{\mathbb{P}_{s,t}, 0 \le s \le t\}$  satisfying the harness property

$$\mathbb{P}_{0,t}\mathbb{F} = (\mathbb{F} + t\mathbb{X})\mathbb{P}_{0,t}$$

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is uniquely determined by the generators  $(\mathbb{A}_t)_{t\geq 0}$ .

### Auxiliary sequence $\mathbb{H}_t \in \mathcal{Q}$

For a polynomial harness  $\{\mathbb{P}_{s,t}\}$ , differentiate wrt *t* 

$$\mathbb{FM}_t = \mathbb{M}_t(\mathbb{F} + t\mathbb{X}), \qquad \mathbb{X} \in \mathcal{Q},$$

to see that

$$\mathbb{A}_t \mathbb{F} - \mathbb{F} \mathbb{A}_t = \mathbb{M}_t \mathbb{X} \mathbb{M}_t^{-1} =: \mathbb{H}_t.$$

**Proposition.**  $\mathbb{H}_t$  uniquely determines  $\mathbb{A}_t$ .

<u>Proof.</u> Since  $\mathbb{A}_t = \mathbb{A}_t \mathbb{FD}$  we have

$$\mathbb{A}_t = \mathbb{F}\mathbb{A}_t\mathbb{D} + \mathbb{H}_t\mathbb{D}.$$

Iterating this we get

$$\mathbb{A}_t = \sum_{k=0}^{\infty} \mathbb{F}^k \mathbb{H}_t \mathbb{D}^{k+1}$$

well defined as element-wise it is a sum of finitely many elements.

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### QH process

**Theorem.** Let { $\mathbb{P}_{s,t}$ ,  $0 \le s \le t$ } be a quadratic harness. Denote  $\mathbb{T}_t = \mathbb{F} - t\mathbb{H}_t$ . Then

$$\mathbb{H}_{t}\mathbb{T}_{t} - \gamma \mathbb{T}_{t}\mathbb{H}_{t} = \mathbb{E} + \theta \mathbb{H}_{t} + \eta \mathbb{T}_{t} + \tau \mathbb{H}_{t}^{2} + \sigma \mathbb{T}_{t}^{2}.$$
 (4)

It follows by the QH property together with

$$\mathbb{H}_t = \mathbb{M}_t \mathbb{X} \mathbb{M}_t^{-1}$$
 and  $\mathbb{T}_t = \mathbb{M}_t \mathbb{F} \mathbb{M}_t^{-1}$ .

lf

$$\gamma \leq \mathbf{1}, \quad \tau, \sigma \leq \mathbf{0}, \quad \text{and} \quad \tau \sigma \neq \mathbf{0}$$

the equation (4) has a unique solution  $\mathbb{H}_t \in \mathcal{Q}$  such that  $h_0 = 0$ .



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### Main goal: Generator $\mathbf{A}_t$ of free QH

The commutation equation has the form

 $(1+\sigma t)\mathbb{H}_{t}\mathbb{F} = \mathbb{E}+\eta\mathbb{F}+\sigma\mathbb{F}^{2}+(\theta-\eta t)\mathbb{H}_{t}-\sigma(t+\tau)\mathbb{F}\mathbb{H}_{t}+(t+\tau)(1+\sigma t)\mathbb{H}_{t}^{2}.$ 

Its solution is

$$\mathbb{H}_t = \frac{1}{1+\sigma t} \left( \mathbb{E} + \eta \mathbb{F} + \sigma \mathbb{F}^2 \right) \phi_t(\mathbb{D}) \mathbb{D},$$

where  $\phi_t(z) = \sum_{k=1}^{\infty} c_k(t) z^{k-1}$  is a power series defined (at least in a neighbourhood of zero) by

$$(z^2 + \eta z + \sigma)(t + \tau)\phi_t^2 + ((\theta - \eta t)z - 2t\sigma - \sigma\tau - 1)\phi_t + t\sigma + 1 = 0$$
  
and  $\phi_t(0) = 1$ .

# Back to the generator

$$\mathbb{A}_{t} = \sum_{k=0}^{\infty} \mathbb{F}^{k} \mathbb{H}_{t} \mathbb{D}^{k+1} = \frac{1}{1+t\sigma} \sum_{k=0}^{\infty} \mathbb{F}^{k} (\mathbb{E} + \eta \mathbb{F} + \sigma \mathbb{F}^{2}) \phi_{t}(\mathbb{D}) \mathbb{D} \mathbb{D}^{k+1}$$
$$= \frac{1}{1+t\sigma} (\mathbb{E} + \eta \mathbb{F} + \sigma \mathbb{F}^{2}) \left( \sum_{k=0}^{\infty} \mathbb{F}^{k} \mathbb{D}^{k+1} \right) \phi_{t}(\mathbb{D}) \mathbb{D}.$$

But

$$\sum_{k=0}^{\infty} \mathbb{F}^k \mathbb{D}^{k+1} = (0, 1, 2x, 3x^2, \dots, kx^{k-1}, \dots) =: \mathbb{D}_1.$$

That is

$$\mathbb{A}_t = \frac{1}{1+t\sigma} \left( \mathbb{E} + \eta \mathbb{F} + \sigma \mathbb{F}^2 \right) \mathbb{D}_1 \phi_t(\mathbb{D}) \mathbb{D}.$$

# Who is $\phi_t$ ?

Let  $G_{\mu}(z) = \int_{\mathbb{R}} \frac{1}{z-x} \mu(dx)$ ,  $z \in \mathbb{C}_+$ , denotes the Cauchy-Stieltjes transform of a measure  $\mu$ . It is well known that  $G_{\mu}$  is an analytic function in  $\mathbb{C}_+$ , determines  $\mu$  uniquely. If it extends to real *z* with |z| large enough then the corresponding moment generating function  $M_{\mu}$  is well defined for |z| small enough and

$$G_{\mu}(z)=rac{1}{z}M_{\mu}(rac{1}{z}).$$

It appears that  $\phi_t(1/z)/z$  agrees with the Cauchy-Stieltjes transform of a probability measure  $\nu_t$  identified in Saitoh and Yosida (2001).

### Operator $\mathbf{H}_t$ through the isomorphism $\Psi$

For  $\mathbf{H}_t$  such that  $\Psi(\mathbf{H}_t) = \mathbb{H}_t$  we have

$$\mathbf{H}_t(x^n) = \frac{1+\eta x+\sigma x^2}{1+\sigma t} \sum_{k=1}^n c_k(t) x^{n-k}.$$

Since  $c_k(t) = \int y^{k-1} v_t(dy)$  we get

$$\mathbf{H}_t(f)(x) = \frac{1+\eta x+\sigma x^2}{1+\sigma t} \int \frac{f(x)-f(y)}{y-x} \nu_t(dy), \qquad f \in \mathcal{P}.$$

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Operator **A**<sub>t</sub>

Since

$$\mathbf{A}_t(x^{n+1}) = \mathbf{H}_t(x^n) + x\mathbf{A}_t(x^n)$$

we get

$$\mathbf{A}_t(f)(x) = \frac{1+\eta x+\sigma x^2}{1+\sigma t} \int \frac{\partial}{\partial x} \left(\frac{f(x)-f(y)}{y-x}\right) \nu_t(dy),$$

 $f \in \mathcal{P}$ .

# Who is $\nu_t$ ?

If 
$$\mu_2(dx) = (ax^2 + bx + c)\mu_1(dx)$$
 and  $\int x\mu_1(dx) = m$ , then  
 $G_{\mu_2}(z) = (az^2 + bz + c)G_{\mu_1}(z) - a(m+z) - b.$  (5)

Let  $\mu_1 = \pi_{t,\eta,\theta,\sigma,\tau}$  be the univariate law of  $X_t$ . Then (5) with

$$m = 0,$$
  $(a, b, c) = \frac{1}{t(t+\tau)}(\tau, \theta t, t^2)$ 

gives

$$\phi_t(z) = G_{\mu_2}(1/z)/z.$$

Therefore,  $\phi_t$  is a moment generating function of the probability measure

$$u_t(\mathbf{dx}) = rac{t^2 + heta t \mathbf{x} + au \mathbf{x}^2}{t(t+ au)} \, \pi_{t,\eta, heta,\sigma, au}(\mathbf{dx}).$$

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