Cauchy-Stielties families with polynomial variance functions

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Abstract

This is a third installment of the series of talks April 3, 2006 and Oct 27, 2011. This talk is based on a forthcoming paper http://arxiv.org/abs/1708.05343 with Raouf Fakhfakh and Wojciech Mlotkowski. The topic of this paper are properties of variance functions of Cauchy-Stieltjes Kernel families generated by a compactly supported (standardized) probability measure. After a brief introduction, I will describe some algebraic operations that can be used to construct additional variance functions from known variance functions. I will describe all quadratic and all cubic variance functions. I will also show how Cauchy-Stieltjes Kernel families with polynomial variance functions are related to generalized orthogonality of some families of polynomials.

Lehmann (1959), Barndorff-Nielsen (1978), Morris (1982), Letac, Mora (1990).

For a probability measure μ on \mathbb{R} we define its *Laplace transform*:

$$L_{\mu}(heta) := \int \exp(heta x) \mu(dx),$$

cumulant function:

$$k_{\mu}(heta) := \log L_{\mu}(heta),$$

and define set

$$\Theta(\mu) := \operatorname{interior} \{\theta : L_{\mu}(\theta) < \infty\}.$$

We assume that μ is not concentrated at one point and that $\Theta(\mu)$ is nonempty.

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For $\theta \in \Theta(\mu)$ we define family of probability distributions

$$P_{\theta}(dx) := \exp(\theta x - k_{\mu}(\theta))\mu(dx)$$

we call this the *natural exponential family* (NEF) generated by μ . The mean function

$$heta\mapsto m(heta)=\int xP_{ heta}(dx)$$

is strictly increasing on $\Theta(\mu)$, so there is an inverse function

$$\psi: M_{\mu} \ni m \mapsto \theta \in \Theta(\mu).$$

The variance function of the NEF generated by μ is defined by

$$V_{\mu}(m) := \int (x-m)^2 P_{\psi(m)}(dx), \quad m \in M_{\mu}.$$

Question: describe variance functions

Morris (1982) characterized those polynomials of order at most 2 which are variance functions. They correspond to Gaussian, Poisson, binomial, negative binomial, gamma and hyperbolic cosine distributions.

Letac and Mora (1990) characterized variance functions which are polynomials of degree at most 3. Here in addition they obtained: Abel, Takacs, strict arcsine, large arcsine, Ressel, inverse Gaussian distribution.



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Kernel family generated by $k(x, \theta)$ and ν

For $\theta \in \Theta(\nu)$,

$$P_{\theta}(dx) = Z_{\theta}^{-1}k(x,\theta)\nu(dx),$$
 where $Z_{\theta} = \int k(x,\theta)\nu(dx)$

Exponential family:

$$k(x, \theta) = \exp(x\theta).$$

Cauchy-Stielties Kernel (CSK) family:

$$k(x,\theta) = \frac{1}{1-x\theta}$$

Cauchy Stielties family generated by a distribution ν :

$$P_{\theta}(dx) := \frac{1}{Z_{\theta}(1-\theta x)} \nu(dx), \qquad \theta \in \Theta, \tag{1}$$

$$Z_{\theta} = \int \frac{1}{1-\theta x}$$

We assume that ν is compactly supported in [a, b], with mean 0 and nondegenerate, with variance, say 1. Then a < 0 < b and $\Theta = (1/a, 1/b)$.

(Somewhat more generally, if ν has mean m_0 then one can find Θ that works, but it is quite convenient to work with the explicit Θ for $m_0 = 0$.)

where

The mean $m(\theta) := \int x P_{\theta}(dx)$ of P_{θ} is

$$m(heta) = rac{Z_{ heta} - 1}{ heta Z_{ heta}}.$$

The function $\theta \mapsto m(\theta)$ is strictly increasing on Θ .

Let $\psi(m)$ denote the inverse function, on a neighborhood of $\int x\nu(dx) = 0$.

The variance function of the Cauchy-Stielties kernel (CSK) family (1) generated by ν is defined by:

$$V(m) := \int (x-m)^2 P_{\psi(m)}(dx).$$

Theorem (Theorem (B-2009))

For m in some neighborhood of $m_0 = 0$ the variance function V(m) is analytic, strictly positive and

$$P_{\psi(m)}(dx) = \frac{V(m)}{V(m) + m(m-x)}\nu(dx).$$

Conversely, if a function V(m) is analytic, strictly positive in a neighborhood of 0 and if ν is a probability distribution, with mean $m_0 = 0$, such that for every m in a neighborhood of 0

$$\frac{V(m)}{V(m) + m(m-x)}\nu(dx)$$

is a probability measure, then ν is compactly supported, non degenerate and determined uniquely.

So we have correspondence:

 $\nu \leftrightarrow V(m),$

where ν is a compactly supported nondegenerate probability distribution with mean 0 and V(m) is an analytic function in a neighborhood of 0, with V(0) > 0.

If $\nu_1 \neq \nu_2$, with the same mean, then the corresponding variance functions $V_1(m)$, $V_2(m)$ are different.

Question:

Which functions V(m) are variance functions?

Notation: ${\cal V}$

The class of all variance functions corresponding to those compactly supported probability distributions ν which have mean 0 and variance 1.

So that if $V \in \mathcal{V}$ then V is an analytic function on a neighborhood of 0 and V(0) = 1.

Therefore the corresponding probability distribution ν has mean 0 and variance 1. We will sometimes write $V = V_{\nu}$.

Notation: \mathcal{V}_{∞}

The class of those $V \in \mathcal{V}$ that the function $m \mapsto V(cm)$ belongs to \mathcal{V} for all real c.

Theorem (B-Ismail-2005)

Suppose $V(m) = 1 + am + bm^2$. Then • $V \in \mathcal{V}$ iff $b \ge -1$ • $V \in \mathcal{V}_{\infty}$ iff $b \ge 0$

Idea of proof

Expand the density of $P_{\psi(m)}(dx) = \frac{V(m)}{V(m)+m(m-x)}\nu(dx)$ into the power series in *m* near m = 0:

$$\frac{V(m)}{V(m)+m(m-x)}=\sum_{k=0}^{\infty}p_k(x)m^k$$

Quadratic variance functions II

The coefficients $p_k(x)$ turn out to be monic polynomials given by the recursion:

$$xp_n(x) = p_{n-1} + \sum_{k=0}^n \frac{V^{(k)}(0)}{k!} p_{n+1-k}(x)$$

For $V(m) = 1 + am + bm^2$, this becomes a "constant" three step recursion

$$xp_n = p_{n+1}(x) + ap_n(x) + (b+1)p_{n-1}(x).$$

Since $P_{\psi(m)}(dx)$ is a probability measure, we have

$$1 = \int_{\mathbb{R}} P_{\psi(m)}(dx) = \int_{\mathbb{R}} \sum_{k=0}^{\infty} p_k(x) m^k \nu(dx) = \sum_{k=0}^{\infty} m^k \int_{\mathbb{R}} p_k(x) \nu(dx)$$
$$= 1 + \sum_{k=1}^{\infty} m^k \int_{\mathbb{R}} p_k(x) \nu(dx)$$

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Thus $\int p_k(x)\nu(dx) = 0$ for k = 1, 2, ... Together with the three-step recursion

$$xp_n = p_{n+1}(x) + ap_n(x) + (b+1)p_{n-1}(x),$$
 (*)

this implies that $\{p_n : n \ge 0\}$ are orthogonal polynomials.

So by Favard's theorem, the last coefficient in the three step recursion (*) must be non-negative: $1 + b \ge 0$.

(Furthermore, b=-1 corresponds to discrete $u=(1-p)\delta_{a}+p\delta_{b}$.)

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Here we write V(m) for the function not for the value!

Theorem

- If $V(m) \in \mathcal{V}$ and $|c| \leq 1$ then $V(cm) \in \mathcal{V}$;
- 3 If $V(m) \in \mathcal{V}$ and $a \in \mathbb{R}$ then $V(m) + am \in \mathcal{V}$;
- 3 If $V(m) \in \mathcal{V}$ then $V(m) + m^2 \in \mathcal{V}_{\infty}$;
- If $V(m) \in \mathcal{V}_{\infty}$ then $V(m) m^2 \in \mathcal{V}$;
- If $V(m) \in \mathcal{V}_{\infty}$ and $a \in \mathbb{R}$ then $V(m) + am \in \mathcal{V}_{\infty}$;
- If $V(m) \in \mathcal{V}_{\infty}$ and $c \geq 1$ then $cV(m) c + 1 \in \mathcal{V}_{\infty}$;
- If $V_1(m), V_2(m) \in \mathcal{V}_{\infty}$ then $V_1(m) + V_2(m) 1 \in \mathcal{V}_{\infty}$.

Corollary (from 3, 4)

The map $V(m) \mapsto V(m) + m^2$ is a bijection of \mathcal{V} onto \mathcal{V}_{∞} .

It turns our this is an old familiar Berkovici-Pata bijection which can be described by a composition of free-boolean and free-additive convolution powers: $\nu \mapsto (\nu^{\boxplus 2})^{\uplus 1/2}$. The later is \boxplus -infinitely divisible by a result of Belinschi-Nica (2008), or because $(\nu^{\boxplus 2})^{\uplus 1/2} = \lim_{\alpha \to \infty} (\nu^{\uplus 1/\alpha})^{\boxplus \alpha}$.

From formulas in a recent manuscript Raouf Fakfhfakh (2019) it follows that $V_{(\nu^{\boxplus 2})^{\uplus 1/2}}(m) = V_{\nu}(m) + m^2$.

Theorem

Consider cubic function $V(m) = 1 + am + bm^2 + cm^3$.

• If
$$(b+1)^3 \ge 27c^2$$
 then $V(m) \in V$.

• $V(m) \in \mathcal{V}_{\infty}$ if and only if $b^3 \ge 27c^2$.

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There is a substantial literature on generalized orthogonality and finite-step recursions for polynomials. We introduce the following generalized orthogonality condition.

Definition

Fix $d \in \mathbb{N}$ and a non-degenerate probability measure ν with moments of all orders. We say that polynomials $\{p_n\}$ are $(\nu; d)$ -orthogonal if

②
$$\int p_n(x)p_k(x)\nu(dx) = 0$$
 for all $n \ge 2 + (k-1)d$, $k = 1, 2...$

Example (d = 1 is the usual orthogonality when ν has ∞ support)

First condition: $p_0 \perp p_1, p_2, \ldots$ Second condition: $p_n \perp p_k$ for $n \ge 2 + (k-1)d = k+1$, so $p_k \perp p_{k+1}, p_{k+2}, \ldots$ for $k = 1, 2, \ldots$

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Example (d = 2)

For d = 2, the conditions are $\int p_n(x)p_0(x)\nu(dx) = 0$ for $n \ge 1$ and $\int p_n(x)p_k(x)\nu(dx) = 0$ for all $n \ge 2 + (k-1)d = 2k$. So $p_0 \perp p_1, p_2, \ldots$ and $p_1 \perp p_2, p_3, \ldots$. But $p_2 \perp p_4, p_5, \ldots$ so there is no restriction on p_2, p_3 . Similarly, $p_3 \perp p_6, p_7, \ldots$ but there are no restrictions on p_3, p_4 and p_3, p_5 .

Somewhat surprisingly, in the setting of Cauchy-Stieltjes kernel families, all the *d*-orthogonality conditions are implied by just two conditions

$$I \quad \int P_n(x)P_0(x)\nu(dx) = 0 \text{ for } n \ge 1,$$

②
$$\int P_n(x)P_2(x)\nu(dx) = 0$$
 for $n \ge 2 + d$,

Question

Does (ν ; d)-orthogonality imply a (d + 2)-step recursion?

The answer to the converse question seems to be quite complicated. Case d = 2 with a perturbation of a constant recursion will be mentioned below. The literature on d + 1-step recursions (Maroni(1989), Iseghem(1987)) assumes non-degeneracy condition that is not satisfied on our case, and the conclusion is the existence of a d-tuple of functionals that need not to be positive.

Generalized orthogonality IV

Theorem

Suppose that $V \in \mathcal{V}$ corresponds to a probability measure ν . Consider

$$\frac{V(m)}{V(m) + m(m-x)} = \sum_{n=0}^{\infty} p_n(x)m^n.$$
 (2)

Then the following statements are equivalent:

- V(m) is a polynomial of degree at most d + 1;
- 2 Polynomials $\{p_n\}$ satisfy (d + 2)-step recursion

$$xp_n(x) = p_{n+1}(x) + \sum_{k=1}^{d+1} b_k p_{n+1-k}(x), \quad n \ge 1$$
 (3)

with initial conditions $p_0(x) = 1$, $p_1(x) = x$, and $p_k \equiv 0$ for k < 0.

- Solution Polynomials $\{p_n(x)\}\$ are $(\nu; d)$ -orthogonal.
- Polynomials {p_n} satisfy conditions in definition of (ν; d)-orthogonality for k = 1, 2 only.

Theorem

Consider polynomials $\{p_n(x)\}\$ given by the (eventually constant) 4-step recursion:

$$\begin{aligned} xp_1(x) &= p_2(x) + ap_1(x) + p_0(x), \\ xp_2(x) &= p_3(x) + ap_2(x) + bp_1(x), \\ xp_n(x) &= p_{n+1}(x) + ap_n(x) + bp_{n-1}(x) + cp_{n-2}(x), \ n \ge 3, \end{aligned}$$

with $p_0(x) = 1$, $p_1(x) = x$. Then the following conditions are equivalent. **1** $b^3 \ge 27c^2$.

 Polynomial {p_n} are (v; 2)-orthogonal for some probability measure v (which then necessarily has mean 0, variance 1, and compact support).

Define
$$\Gamma_1(p) = \int_{\mathbb{R}} p(x)\nu(dx)$$
 and $\Gamma_2(p) = \int_{\mathbb{R}} xp(x)\nu(dx)$.

If polynomials $p_n(x)$ are $(\nu; d)$ -orthogonal with d = 2. Then $\Gamma_{\alpha}(p_n p_k) = 0$ if $n \ge 2k + \alpha$ for $\alpha = 1, ..., d$; however the non-singularity part of the requirement of d = 2-orthogonality that $\Gamma_{\alpha}(p_n p_k) \ne 0$ if $n = 2k + \alpha - 1$ fails.

Similarly, for arbitrary d, define $\Gamma_j(p) = \int x^{j-1} p(x) \nu(dx)$. Then $\Gamma_\alpha(p_n p_k) = 0$ if $n \ge dk + \alpha$, but $(\nu; d)$ -orthogonality yields more zeros than what is allowed under d-orthogonality.

Thank you!

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KERNEL FAMILIES

Let μ be a positive measure on $(\mathbf{R}, \mathcal{B}(\mathbf{R}))$, and $k : \mathbf{R}^2 \to \mathbf{R}_+$ a measurable function, which will play the role of the kernel. Denote

$$G_{\mu}(\theta) = \int_{\mathbf{R}} k(x,\theta) \,\mu(dx),$$

for such θ that $\oint_{\mathbb{R}} k(x, \theta) \ \mu dx < \infty$. Denote additionally

$$\Theta = Int\{\theta \in \mathbf{R} : \int_{\mathbf{R}} k(x, \theta) \ \mu(dx) < \infty\}.$$

The natural exponential family is an example of a kernel family with $k(x, \theta) = e^{\theta x}$. Consider two new special cases:

GAMMA KERNEL FAMILY



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