

The Covariance Representation on the Sphere

Sergey Bobkov
University of Minnesota

joint work with Devraj Duggal

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One dimensional covariance identities

Höfdding (1940) Given random variables X and Y with finite second moments, we have

$$\begin{aligned}\text{cov}(X, Y) &= \mathbb{E}XY - \mathbb{E}X \mathbb{E}Y \\ &= \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} H(x, y) dx dy,\end{aligned}$$

where the Höfdding kernel is defined by

$$H(x, y) = \mathbb{P}\{X \leq x, Y \leq y\} - \mathbb{P}\{X \leq x\} \mathbb{P}\{Y \leq y\}.$$

Generalized Höfdding formula (Mardia, Sen, Cuadras ...)

$$\text{cov}(u(X), v(Y)) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} u'(x)v'(y) H(x, y) dx dy.$$

The case $X = Y$ with distribution μ

Theorem. Given a probability measure μ on the real line,

$$\text{cov}_\mu(u, v) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} u'(x)v'(y) d\lambda(x, y)$$

for a unique positive, locally finite measure λ , with density

$$H(x, y) = \frac{d\lambda(x, y)}{dx dy} = F(x \wedge y) (1 - F(x \vee y)),$$

where

$$F(x) = \mathbb{P}\{X \leq x\} = \mu((-\infty, x]),$$

$$x \wedge y = \min(x, y), \quad x \vee y = \max(x, y).$$

Definition: λ is called Höfding's measure.

Assumptions:

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |u'(x)| |u'(y)| d\lambda(x, y) < \infty,$$

$$\int_{-\infty}^{\infty} \int_{-\infty}^{\infty} |v'(x)| |v'(y)| d\lambda(x, y) < \infty.$$

Total mass:

$$\lambda(\mathbb{R} \times \mathbb{R}) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} H(x, y) dx dy = \text{Var}(X).$$

Marginals of Höfding measures

Let μ have finite second moment with $a = \mathbb{E}X$. The measure λ has equal unimodal marginals: For Borel sets $A \subset \mathbb{R}$,

$$\Lambda(A) = \lambda(A \times \mathbb{R}) = \int_A \int_{-\infty}^{\infty} H(x, y) dx dy$$

with density

$$h(x) = \frac{d\Lambda(x)}{dx} = \int_x^{\infty} (y - a) dF(y).$$

Corollary.

$$\text{cov}(X, u(X)) = \int_{-\infty}^{\infty} u'(x) h(x) dx.$$

If μ is supported on an interval Δ and has there a positive density p , this formula may be rewritten as Stein's identity

$$\text{cov}(X, u(X)) = \mathbb{E} \tau(X) u'(X),$$

where

$$\tau(x) = \frac{h(x)}{p(x)} = \frac{1}{p(x)} \int_x^{\infty} (y - a) p(y) dy, \quad x \in \Delta,$$

is the Stein kernel.

Characterization of Gaussian measures via marginals of Höfding measures

Claim. $\Lambda = \sigma^2\mu$ for some $\sigma \geq 0$ iff $\mu \sim N(a, \sigma^2)$.

Proof. If $a = 0$, the Fourier-Stieltjes transform of Λ is

$$\hat{\Lambda}(t) = \int_{-\infty}^{\infty} e^{itx} h(x) dx = -\frac{f'(t)}{t} \quad (t \neq 0),$$

where f is the characteristic function of X . Hence $\Lambda = \sigma^2\mu$ iff

$$f'(t) = -\sigma^2 t f(t) \quad \text{for all } t \in \mathbb{R},$$

that is, $\mu \sim N(0, \sigma^2)$.

Standard normal case. If $a = 0$, $\sigma = 1$, we have Stein's identity with $\tau = 1$,

$$\text{cov}(X, u(X)) = \mathbb{E} u'(X).$$

Multidimensional extension

Let μ be the standard Gaussian measure on \mathbb{R}^n with density

$$\varphi(x) = \frac{1}{(2\pi)^{n/2}} e^{-|x|^2/2}, \quad x \in \mathbb{R}^n.$$

Theorem. For a (unique) probability measure λ on $\mathbb{R}^n \times \mathbb{R}^n$,

$$\text{cov}_\mu(u, v) = \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} \langle \nabla u(x), \nabla v(y) \rangle d\lambda(x, y).$$

Necessarily $\Lambda = \mu$.

History: Via Ornstein-Uhlenbeck semigroups (Ledoux 1995):

$$\text{cov}_\mu(u, v) = \int_0^\infty \mathbb{E}_\mu \langle \nabla u, \nabla P_t v \rangle dt,$$

$$P_t v(x) = \int_{\mathbb{R}^n} v(e^{-t}x + \sqrt{1 - e^{-2t}}y) d\mu(y).$$

Interpolation (Houdré, Pérez-Abreu 1995, H-PA-Surgailis 1998),
B-Houdré-Götze (2001)

Uniqueness of λ : Apply the identity to

$$u(x) = e^{i\langle a, x \rangle}, \quad v(y) = e^{i\langle b, y \rangle} \quad (a, b \in \mathbb{R}^n).$$

Characterization of Gaussian measures via covariance identity

Theorem. Given a probability measure μ on \mathbb{R}^n , suppose that

$$\text{cov}_\mu(u, v) = \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} \langle \nabla u(x), \nabla v(y) \rangle d\lambda(x, y)$$

for some finite measure λ on $\mathbb{R}^n \times \mathbb{R}^n$ in the class of all bounded smooth u, v on \mathbb{R}^n with bounded partial derivatives. Then μ is Gaussian with covariance matrix $\sigma^2 I_n$.

Proof is based on the Darmois-Skitovitch theorem.

Gaussian concentration

Let μ be the standard Gaussian measure on \mathbb{R}^n .

Let u be a Lipschitz function on \mathbb{R}^n with $\|u\|_{\text{Lip}} \leq 1$ and mean $m = \mathbb{E}u$.

Log-Sobolev: For all $r > 0$,

$$\mu\{|u - m| \geq r\} \leq 2e^{-r^2/2}.$$

Isoperimetry:

$$\mu\{|u - m| \geq r\} \leq 4 \frac{e^{-r^2/2}}{r}.$$

Corollary (from the covariance identity):

$$\mu\{|u - m| \geq r\} \leq \mathbb{E} |u - m| \frac{e^{-r^2/2}}{r}.$$

Example: $u(x) = \max(x_1, \dots, x_n)$.

Proof

Let $m = 0$ and $v = T(u)$ with non-decreasing T . Then

$$\text{cov}_\mu(u, T(u)) = \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} \langle \nabla u(x), \nabla u(y) \rangle T'(u(y)) d\lambda(x, y)$$

implies

$$\mathbb{E} uT(u) \leq \mathbb{E} T'(u).$$

Choose

$$T(x) = \min\{(x - r)^+, \varepsilon\}, \quad x, \varepsilon > 0.$$

Let $\varepsilon \rightarrow 0$, we get

$$V(r) \equiv \int_r^\infty xp(x) dx \leq p(r),$$

where p is density of u under μ . Equivalently, $V(r) e^{r^2/2}$ is non-increasing in $r > 0$, so,

$$V(r) e^{r^2/2} \leq V(0) = \mathbb{E} u^+.$$

But $V(r) \geq r\mu\{u \geq r\}$, and we arrive at

$$\mu\{u \geq r\} \leq \mathbb{E} u^+ \frac{e^{-r^2/2}}{r}.$$

Spherical derivatives

Let $\mu = \sigma_{n-1}$ be a uniform distribution on the unit sphere

$$S^{n-1} = \{x \in \mathbb{R}^n : |x| = 1\}, \quad n \geq 2.$$

If f is smooth on the sphere, then $w = \nabla_S f(x)$ at the point $x \in S^{n-1}$ is the shortest vector such that

$$f(y) = f(x) + \langle w, y - x \rangle + o(|y - x|)$$

for $y \rightarrow x$, $y \in S^{n-1}$. If f is smooth in a neighborhood of the sphere,

$$\nabla_S f(x) = P_{x^\perp} \nabla f(x) = \nabla f(x) - \langle \nabla f(x), x \rangle x.$$

Spherical covariance identity

Theorem. On $S^{n-1} \times S^{n-1}$ there exist a probability measure ν with marginals σ_{n-1} and a constant $c_n > 0$ such that

$$\text{COV}_{\sigma_{n-1}}(f, g) = c_n \int_{S^{n-1}} \int_{S^{n-1}} \langle \nabla_S f(x), \nabla_S g(y) \rangle d\nu(x, y)$$

for all smooth f, g on S^{n-1} . Moreover, ν has density with respect to $\sigma_{n-1} \otimes \sigma_{n-1}$ of the form $\psi(\langle x, y \rangle)$. We also have

$$\frac{1}{n-1} < c_n < \frac{1}{n-2} \quad (n \geq 3).$$

Notes. 1) Since $\langle x, y \rangle = 1 - \frac{1}{2}|x - y|^2$, the density depends on the distance $|x - y|$.

2) In general no uniqueness. If $S \subset S^{n-1}$ is a circle, and ν' is supported on the set $A = \{(x, y) \in S \times S : \langle x, y \rangle = 0\}$, then

$$\langle \nabla_S f(x), \nabla_S g(y) \rangle = 0 \quad \text{for } (x, y) \in A.$$

Hence, a similar covariance identity is also true for $c_n \nu + \nu'$.

Density in Gaussian covariance representation

The probability measure λ in

$$\text{cov}_\mu(u, v) = \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} \langle \nabla u(x), \nabla v(y) \rangle d\lambda(x, y)$$

may be described as

$$\lambda = \int_0^1 \mathcal{L}(X, tX + sZ) dt, \quad s = \sqrt{1 - t^2}.$$

Here X and Z are independent standard normal random vectors in \mathbb{R}^n . It has density on $\mathbb{R}^n \times \mathbb{R}^n$

$$p(x, y) = \frac{1}{(2\pi)^n} \int_0^1 s^{-n} \exp \left[-\frac{|x|^2 + |y|^2 - 2t \langle x, y \rangle}{2s^2} \right] dt.$$

Construction of spherical mixing density

With smooth $f, g : S^{n-1} \rightarrow \mathbb{R}$, we associate the functions

$$\begin{aligned} u(x) &= f(r^{-1}x) = f(\theta), & r &= |x|, & \theta &= r^{-1}x, \\ v(x) &= g(r^{-1}x) = g(\theta), \end{aligned}$$

on $\mathbb{R}^n \setminus \{0\}$, with their gradients

$$\nabla u(x) = \frac{1}{r} \nabla_S f(\theta), \quad \nabla v(y) = \frac{1}{r'} \nabla_S g(\theta')$$

Here r and θ are independent under μ , and $\mathcal{L}_\mu(u) = \mathcal{L}_{\sigma_{n-1}}(f)$. Hence

$$\text{cov}_{\sigma_{n-1}}(f, g) = \text{cov}_\mu(u, v).$$

Applying the Gaussian covariance identity and integrating in polar coordinates, we represent $\text{cov}_{\sigma_{n-1}}(f, g)$ as

$$\iint \langle \nabla_S f(\theta), \nabla_S g(\theta') \rangle \psi(\langle \theta, \theta' \rangle) d\sigma_{n-1}(\theta) d\sigma_{n-1}(\theta'),$$

$$\begin{aligned} \psi(\alpha) &= \frac{1}{2^{n-2} \Gamma(\frac{n}{2})^2} \int_0^1 s^{n-2} \left[\int_0^\infty \int_0^\infty \right. \\ &\quad \left. \exp \left[-\frac{r^2 + r'^2 - 2rr't\alpha}{2} \right] (rr')^{n-2} dr dr' \right] dt. \end{aligned}$$

Marginals. Estimation of constants c_n

Repeating integration in polar coordinates, we get

$$\begin{aligned} c_n &= \iint \psi(\langle \theta, \theta' \rangle) d\sigma_{n-1}(\theta) d\sigma_{n-1}(\theta') \\ &= \int_{\mathbb{R}^n} \int_{\mathbb{R}^n} \frac{p(x, y)}{|x| |y|} dx dy = \mathbb{E} \int_0^1 \frac{1}{|X| |tX + sZ|} dt. \end{aligned}$$

By Cauchy's inequality, the latter expectation is smaller than

$$\left(\mathbb{E} \frac{1}{|X|^2} \right)^{1/2} \left(\mathbb{E} \frac{1}{|tX + sZ|^2} \right)^{1/2} = \mathbb{E} \frac{1}{|X|^2} = \frac{1}{n-2}.$$

Rotational invariance: The density of the marginal of ν is

$$q(\theta) = \int_{S^{n-1}} \psi(\langle \theta, \theta' \rangle) d\sigma_{n-1}(\theta').$$

This integral does not depend on θ and is therefore equal to c_n .

Behavior of ψ near the diagonal

Write $\psi = \psi_n$.

Claim: ψ_2 is bounded, while (within n -dependent factors)

$$\psi_3(\langle \theta, \theta' \rangle) \sim \log \frac{1}{|\theta - \theta'|} \quad (n = 3),$$

$$\psi_n(\langle \theta, \theta' \rangle) \sim \frac{1}{|\theta - \theta'|^{n-3}} \quad (n \geq 4).$$

Application to spherical concentration

Let f on S^{n-1} satisfy $\|f\|_{\text{Lip}} \leq 1$ and have σ_{n-1} -mean m .

Classical deviation inequality: For all $r > 0$,

$$\sigma_{n-1} \left\{ |f - m| \geq r \right\} \leq 2 e^{-(n-1)r^2/2}.$$

Corollary. For all $r > 0$,

$$\sigma_{n-1} \left\{ |f - m| \geq r \right\} \leq \frac{1}{r} e^{-r^2/2c_n} \mathbb{E}_{\sigma_{n-1}} |f - m|.$$

Periodic covariance identities

For a probability measure μ on $[0, 1)$, consider an identity

$$\text{cov}_\mu(u, v) = \int_0^1 \int_0^1 u'(x)v'(y) d\lambda(x, y)$$

in the class of all 1-periodic smooth functions u and v on \mathbb{R} , assuming that λ is a signed measure.

Example: The Höfdding measure $\lambda = \lambda_\mu$ whose marginal Λ_μ are however not multiples of μ .

Theorem. Subject to the constraint that the marginal distribution of λ is equal to $c\mu$ for a prescribed value $c \in \mathbb{R}$, the mixing measure λ exists, is unique and is given by

$$\begin{aligned} \lambda = & \lambda_\mu + (\sigma^2 - c) m \otimes m \\ & + c(\mu \otimes m + m \otimes \mu) - (\Lambda_\mu \otimes m + m \otimes \Lambda_\mu), \end{aligned}$$

where σ^2 is variance of μ and m is the uniform distribution on $(0, 1)$.

Uniform distribution

Let μ be uniform on $(0, 1)$.

Corollary. Subject to the constraint that the marginal distribution of λ is equal to $c\mu$, $c \in \mathbb{R}$, the measure λ has density

$$\frac{\lambda(x, y)}{dx dy} = D(|x - y|) + \left(c - \frac{1}{24}\right), \quad x, y \in (0, 1),$$

where

$$D(h) = \frac{1}{8} [1 - 4h(1 - h)], \quad 0 \leq h \leq 1.$$

Optimal choice $c = \frac{1}{24}$ leads to

$$\begin{aligned} \text{cov}_\mu(u, v) &= \int_0^1 \int_0^1 u'(x)v'(y) D(|x - y|) dx dy \\ &= \frac{1}{24} \int_0^1 \int_0^1 u'(x)v'(y) d\nu(x, y) \end{aligned}$$

with a probability measure $d\nu(x, y) = 24 D(|x - y|) dx dy$ on $(0, 1) \times (0, 1)$. It has μ as a marginal one.