Random (Beta) Polytopes

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Based on joint works with Florian Besau, Thomas Godland, Anna Gusakova, Zakhar Kabluchko, Olexander Marynych, Matthias Reitzner, Daniel Rosen, Carsten Schütt, Daniel Temesvari, Elisabeth Werner

Classical random polytopes

• $K \subset \mathbb{R}^d$ a convex body (compact, convex & non-empty interior)



• Two classical models: For $n \in \mathbb{N}$ let

• X_1, \ldots, X_n be uniformly distributed in the interior of K

Random polytope $[K]_n = \operatorname{conv}\{X_1, ..., X_n\}$

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• $X_1, ..., X_n$ be uniformly distributed on the boundary of KRandom inscribed polytope $[\partial K]_n = \operatorname{conv}\{X_1, ..., X_n\}$



- Rényi & Sulanke (1963) d = 2, K smooth enough
- Schneider & Wieacker (1980) $d \ge 2, K$ smooth enough
- Bárány (1982) $d \ge 2, K$ of class C_+^3
- Schütt (1994), Böröczky, Fodor, Hug (2010) $d \ge 2, K$ general





• The constant as(K) is the affine surface area of *K*:

$$\operatorname{as}(K) = \int_{\partial K} H_{d-1}(K; x)^{\frac{1}{d+1}} \mu_K(\mathrm{d}x).$$



- Blaschke (1923), Santaló (1949) d = 2, d = 3, K smooth enough
- Leichtweiss (1988), Schütt & Werner (1990) $d \ge 2$, general K
- Ludwig & Reitzner (1999) characterization

Theorem (Reitzner 2003) Suppose *K* is of class C_+^2 . Then $c_1(K)n^{-1-\frac{2}{d+1}} \leq \operatorname{Var}\operatorname{Vol}([K]_n) \leq c_2(K)n^{-1-\frac{2}{d+1}}$

- Küfer (1994) upper bound if $K = B^d$
- Buchta (2005) lower bound d = 2, K smooth enough
- Calka & Yukich (2014) asymptotics, K smooth enough



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Theorem (Reitzner 2005)
Suppose *K* is of class
$$C_{+}^{2}$$
. Then $\frac{\operatorname{Vol}([K]_{n}) - \mathbb{E}\operatorname{Vol}([K]_{n})}{\sqrt{\operatorname{Var}\operatorname{Vol}([K]_{n})}} \xrightarrow{D} Z \sim \mathcal{N}(0,1)$

Theorem

$$\operatorname{EVol}([\partial K]_n) = \operatorname{Vol}(K) - \tilde{c}(d)\Omega(K) \left(\frac{\mu_K(K)}{n}\right)^{\frac{2}{d-1}} (1 + o_n(1))$$

Buchta, Müller & Tichy (1985): K of class C_+^3 Reitzner (2002), Schütt & Werner (2001): K of class C_+^2

Suppose *K* is of class C_+^2 . Then

$$c_1(K)n^{-1-\frac{4}{d-1}} \le \operatorname{Var}\operatorname{Vol}([\partial K]_n) \le c_2(K)n^{-1-\frac{4}{d-1}}$$

Richardson, Vu, Wu (2008)

 $\frac{\operatorname{Vol}([\partial K]_n) - \mathbb{E}\operatorname{Vol}([\partial K]_n)}{\sqrt{\operatorname{Var}\operatorname{Vol}([\partial K]_n)}} \xrightarrow{D} Z \sim \mathcal{N}(0,1)$

T. (2018)



Theorem





Random polytopes on the sphere

• $K \subset \mathbb{S}^d$ a spherical convex body (geodesically convex, contained in open hemisphere)



- Two classical models: For $n \in \mathbb{N}$ let
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Theorem

Suppose *K* is of class C_+^2 . Then

$$\mathbb{E}\operatorname{Vol}_{s}([K]_{n}) = \operatorname{Vol}_{s}(K) - c(d)\operatorname{as}_{s}(K) \left(\frac{\operatorname{Vol}_{s}(K)}{n}\right)^{\frac{2}{d+1}} (1 + o_{n}(1))$$

Besau, Ludwig, Werner (2018)

$$\mathbb{E}\operatorname{Vol}_{s}([\partial K]_{n}) = \operatorname{Vol}_{s}(K) - \widetilde{c}(d)\Omega_{s}(K) \left(\frac{\operatorname{Vol}_{s}(K)}{n}\right)^{\frac{2}{d-1}} (1 + o_{n}(1))$$

Follows from Böröczky, Fodor, Hug (2013) and Besau, Ludwig, Werner (2018)



Theorem

Suppose K is of class C_{+}^{2} . Then $\operatorname{Var}\operatorname{Vol}_{s}([K]_{n}) \geq c(K)n^{-1-\frac{2}{d+1}}$ Besau & Thäle (2020) $\frac{\operatorname{Vol}_{s}([K]_{n}) - \mathbb{E}\operatorname{Vol}_{s}([K]_{n})}{\sqrt{\operatorname{Var}\operatorname{Vol}_{s}([K]_{n})}} \xrightarrow{D} Z \sim \mathcal{N}(0,1)$ $\operatorname{Var}\operatorname{Vol}_{s}([\partial K]_{n}) \geq c(K)n^{-1-\frac{4}{d-1}}$ Besau, Rosen & Thäle (2021) $\frac{\operatorname{Vol}_{s}([\partial K]_{n}) - \mathbb{E}\operatorname{Vol}_{s}([\partial K]_{n})}{\sqrt{\operatorname{Var}\operatorname{Vol}_{s}([\partial K]_{n})}} \xrightarrow{D} Z \sim \mathcal{N}(0,1)$





- The proof for asymptotic normality relies on a version of Stein's method due to Chatterjee (2008) as presented by Lachiéze-Rey & Peccati (2017),
- geometric estimates involving weighted floating bodies and weighted surface bodies,
- and a projection argument reducing the problem to \mathbb{R}^d .

The projection argument



$$X_i \sim \operatorname{Vol}_s|_K$$
$$[K]_n = \operatorname{conv}_s \{X_1, \dots, X_n\}$$
$$\operatorname{Vol}_s([K]_n)$$

 $\bar{X}_i \sim (g \sharp \operatorname{Vol}_s) |_{\bar{K}}$ $[\bar{K}]_n = \operatorname{conv} \{ \bar{X}_1, \dots, \bar{X}_n \}$ $\operatorname{Vol}_{\varphi}([\bar{K}]_n) = \int_{[\bar{K}]_n} \varphi(x) \, \mathrm{d}x$

 $\varphi = \text{density of } g \sharp \text{Vol}_s$

Stein's method

 $K \subset \mathbb{R}^d$ a convex body, X_1, \ldots, X_n independent random points in K distributed according to some *nice* density ψ . Consider the weighted volume $\operatorname{Vol}_{\varphi}([K]_n^{\psi})$ where φ is another *nice* density function on K.

$$d_{\text{Wass}}(V_n, Z) \le \frac{c}{\text{Var Vol}_{\varphi}([K]_n^{\psi})} \left(\gamma_1 + \gamma_2 + \gamma_3 + \gamma_4\right)$$

where γ_1 , γ_2 , γ_3 , γ_4 are terms only involving the first and second order difference operator applied to the weighted volume functional:

$$D_1 \operatorname{Vol}_{\varphi}([K]_n^{\psi}) = \operatorname{Vol}_{\varphi}(\operatorname{conv}\{X_1, \dots, X_n\}) - \operatorname{Vol}_{\varphi}(\operatorname{conv}\{X_2, \dots, X_n\})$$
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Especially $\gamma_3 \simeq \mathbb{E} |D_1 \operatorname{Vol}_{\varphi}([K]_n^{\psi})|^3$ and $\gamma_4 \simeq \mathbb{E} |D_1 \operatorname{Vol}_{\varphi}([K]_n^{\psi})|^4$.

The weighted floating body of K is defined by

$$K^{\varphi}_{\delta} = \bigcap \left\{ H^{-} : \operatorname{Vol}_{\varphi}(K \cap H^{+}) \leq \delta \right\}$$

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$$[K]_n^{\psi}$$
 contains $K_{c_1 \frac{\log n}{n}}^{\varphi}$ with
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 $\gamma_4 \asymp \mathbb{E} \left| D_1 \operatorname{Vol}_{\varphi}([K]_n^{\psi}) \right|^4$

 $\leq \mathbb{E}[|D_1 \operatorname{Vol}_{\varphi}([K]_n^{\psi})|^4 |A_n] + c_3 \mathbb{P}(A_n^c)$







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$$\begin{split} \gamma_4 &\asymp \mathbb{E} \left| D_1 \operatorname{Vol}_{\varphi}([K]_n^{\psi}) \right|^4 \\ &\leq \mathbb{E} \left[\left| D_1 \operatorname{Vol}_{\varphi}([K]_n^{\psi}) \right|^4 \left| A_n \right] + c_3 \mathbb{P}(A_n^c) \\ &\leq \mathbb{E} \operatorname{Vol}_{\psi} \left(\Delta_K^{\varphi} \left(X_1, c_1 \frac{\log n}{n} \right) \right)^4 + c_4 n^{-7} \end{split}$$





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$$\leq c_5 \left(\frac{\log n}{n}\right)^4 \operatorname{Vol}_{\psi}(K \setminus K_{c_1 \frac{\log n}{n}}^{\varphi}) + c_4 n^{-7}$$





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$$\leq c_6 \left(\frac{\log n}{n}\right)^{4 + \frac{2}{d+1}}$$





Random polytopes in hyperbolic space



Suppose *K* is of class C_+^2 . Then

$$\mathbb{E}\operatorname{Vol}_{h}([K]_{n}) = \operatorname{Vol}_{h}(K) - c(d)\operatorname{as}_{h}(K) \left(\frac{\operatorname{Vol}_{h}(K)}{n}\right)^{\frac{2}{d+1}} (1 + o_{n}(1))$$

 $\operatorname{Var}\operatorname{Vol}_{h}([K]_{n}) \geq c(K)n^{-1-\frac{2}{d+1}}$

$$\frac{\operatorname{Vol}_{h}([K]_{n}) - \mathbb{E}\operatorname{Vol}_{h}([K]_{n})}{\sqrt{\operatorname{Var}\operatorname{Vol}_{h}([K]_{n})}} \xrightarrow{D} Z \sim \mathcal{N}(0,1)$$

Besau, Ludwig, Werner (2018)

Besau & Thäle (2020)





Suppose *K* is of class C_+^2 .

$$\mathbb{E}\operatorname{Vol}_{s}([K]_{n}) = \operatorname{Vol}_{s}(K) - c(d, K) \left(\frac{\operatorname{Vol}_{s}(K)}{n}\right)^{\frac{2}{d+1}} (1 + o_{n}(1))$$



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... random polytopes in a spherical wedge

$$\mathbb{E}\operatorname{Vol}_{s}([K]_{n}) = \operatorname{Vol}_{s}(K) - c(d)\frac{\log n}{n}(1 + o_{n}(1))$$

Besau, Gusakova, Reitzner, Schütt, T. & Werner (2022)

In this case g(K) = a strip and $g \# Vol_s$ has some density.







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In this case g(K) = a strip and $g \# Vol_s$ has some density.





Suppose K is the half-sphere.

 \mathbb{E} Vol_s([K]_n) = Vol_s(K) - c(d)n⁻¹(1 + o_n(1))

Bárány, Hug, Reitzner & Schneider (2017)

In this case $g(K) = \mathbb{R}^d$ and $g \sharp \operatorname{Vol}_s$ has density $\propto (1 - \|x\|^2)^{-\frac{d+1}{2}}$.

$$f_{d,\beta}(x) = c_{d,\beta}(1 - \|x\|^2)^{\beta} \qquad (\|x\| \le 1, \beta > -1)$$

• Beta polytope: $P_{n,d}^{\beta} = [X_1, ..., X_n]$

 X_1, \ldots, X_n iid with density $f_{d,\beta}$

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• Special cases: $\beta \rightarrow -1$ uniform distribution on \mathbb{S}^{d-1}

There are weak sub-sequential limits by Helly's compactness theorem. All sub-sequential limits must be concentrated on \mathbb{S}^{d-1} . All sub-sequential limits must be rotationally invariant.

 $\beta = 0$ uniform distribution on \mathbb{B}^d

 $\beta \to \infty$ Gaussian distribution on \mathbb{R}^d

$$X \sim f_{d,\beta} \Longrightarrow \sqrt{2\beta} X \sim c_{d,\beta} \left(1 - \frac{\|x\|^2}{2\beta}\right)^{\beta} \longrightarrow c e^{-\|x\|^2/2}$$

Now, apply Scheffé's lemma.

$$f_{d,\beta}(x) = c_{d,\beta}(1 - \|x\|^2)^{\beta} \qquad (\|x\| \le 1, \beta > -1)$$

• Beta prime distribution in \mathbb{R}^d :

$$\tilde{f}_{d,\beta}(x) = \tilde{c}_{d,\beta}(1 + \|x\|^2)^{-\beta} \qquad (x \in \mathbb{R}^d, \beta > d/2)$$

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• Poisson polytope:

$$\eta = \operatorname{PPP}\left(\frac{\mathrm{d}x}{\|x\|^{d+\alpha}}\right), \qquad \alpha > 0$$

 $\Pi_{d,\alpha} = [\eta]$

Poisson polytopes are rescaled limits of beta prime polytopes:

$$n^{-1}\tilde{P}^{\beta}_{n,d} \longrightarrow \Pi_{d,2\beta-d}$$



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• Key properties:

Projection invariance



$$f_{d,\beta}(x) = c_{d,\beta}(1 - \|x\|^2)^{\beta} \qquad (\|x\| \le 1, \beta > -1)$$

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• Known results:

 $\mathbb{E}f_{k}(P_{n,d}^{\beta}), \quad \mathbb{E}f_{k}(\tilde{P}_{n,d}^{\beta})$ $\mathbb{E}V_{k}(P_{n,d}^{\beta}), \quad \mathbb{E}V_{k}(\tilde{P}_{n,d}^{\beta})$

Angles of tangent cones

Explicit formulas Asymptotic expansions Threshold phenomena

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Application to stochastic geometry models



 $\mathbb{S}^d_+ := \{ (x_0, x_1, \dots, x_d) \in \mathbb{R}^{d+1} : x_0^2 + x_1^2 + \dots + x_d^2 = 1, x_0 \ge 0 \}$

upper half-sphere in \mathbb{R}^{d+1}

 U_1, U_2, \dots iid uniform random points on \mathbb{S}^d_+

$$C_n := \operatorname{pos}(U_1, \dots, U_n) = \left\{ \sum_{i=1}^n \lambda_i U_i : \lambda_i \ge 0 \right\}$$

random cone generated by $U_1, ..., U_n$



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random cone generated by $U_1, ..., U_n$



$$(x_0, x_1, \dots, x_d) \mapsto \left(\frac{x_1}{x_0}, \dots, \frac{x_d}{x_0}\right)$$
$$U_i \mapsto V_i$$

Density of
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: $\frac{c_d}{(1 + \|x\|^2)^{\frac{d+1}{2}}} = \tilde{f}_{d,\frac{d+1}{2}}$



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Density of
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$$(x_0, x_1, \dots, x_d) \mapsto \left(\frac{x_1}{x_0}, \dots, \frac{x_d}{x_0}\right)$$
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Density of
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Take $B \subset \mathbb{R}^d \setminus \{0\}$: $\sum_{i=1}^n \mathbf{1} \left\{ \frac{V_i}{n} \in B \right\} \sim \operatorname{Bin}\left(n, \frac{c_d}{n} \int_B \frac{dx}{\|x\|^{d+1}}\right) \xrightarrow{d} \operatorname{Po}\left(c_d \int_B \frac{dx}{\|x\|^{d+1}}\right)$



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 $\implies \sum_{i=1}^n \delta_{\frac{V_i}{n}} \xrightarrow{w} \operatorname{PPP}\Big(\frac{c_d dx}{\|x\|^{d+1}}\Big)$



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$$\begin{array}{lll} \text{Density of } \displaystyle \frac{V_i}{n}: & c_d \frac{n^d}{(1+\|nx\|^2)^{\frac{d+1}{2}}} = c_d \frac{n^d}{n^{d+1}(\frac{1}{n^2}+\|x\|^2)^{\frac{d+1}{2}}} \sim \frac{1}{n} \frac{c_d}{\|x\|^{d+1}} \\ \text{Take } B \subset \mathbb{R}^d \backslash \{0\}: & \displaystyle \sum_{i=1}^n \mathbf{1} \Big\{ \frac{V_i}{n} \in B \Big\} \sim \text{Bin}\Big(n, \frac{c_d}{n} \int_B \frac{dx}{\|x\|^{d+1}}\Big) \xrightarrow{d} \text{Po}\Big(c_d \int_B \frac{dx}{\|x\|^{d+1}}\Big) \\ & \Longrightarrow \sum_{i=1}^n \delta_{\frac{V_i}{n}} \xrightarrow{w} \text{PPP}\Big(\frac{c_d dx}{\|x\|^{d+1}}\Big) \\ & \Longrightarrow \frac{1}{n} [V_1, \dots, V_n] \xrightarrow{w} \Big[\text{PPP}\Big(\frac{c_d dx}{\|x\|^{d+1}}\Big) \Big] & \swarrow \\ & \Longrightarrow \mathbb{E} f_{k+1}(C_n) = \mathbb{E} f_k(n^{-1}[V_1, \dots, V_n]) \longrightarrow \mathbb{E} f_k(\Pi_{d,1}) \\ & \text{Example } d = 2: & \mathbb{E} f_0(C_n) \rightarrow \mathbb{E} f_0(\Pi_{d,1}) = \frac{\pi^2}{2} \end{array}$$

Application 2: Typical Voronoi cells on the sphere

- $X_1, ..., X_n$ iid uniform random points on \mathbb{S}^d
- $C_i = \{z \in \mathbb{S}^d : d_g(z, X_i) \le d_g(z, X_j) \; \forall j\}$

Voronoi cell of X_i

- $\{C_1, ..., C_n\}$ Voronoi tessellation
- V_n Typical Voronoi cell = cell picked uniformly at random



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 η_t stationary Poisson point process in \mathbb{H}^d

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Theorem: $\mathbb{E}f_k(V_0)$ can be computed explicitly in terms of a beta-star polytope.



Thank you!

R.S