From statistics to topology and back again

Robert Adler

Industrial Engineering & Management and Electrical Engineering

Technion - Israel Institute of Technology

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Back to the future

Early Development of the Universe **BIG BANG BIG BANG PLUS TINIEST** FRACTION OF A SECOND (10-43) INFLATION **BIG BANG PLUS** 300.000 YEARS LIGHT FROM FIRST GALAXIES BIG BANG PLUS 15 BILLION YEARS



The Nobel Prize in Physics 2006

"for their discovery of the blackbody form and anisotropy of the cosmic microwave background radiation"



John C. Mather NASA b. 1946



George F. Smoot Berkeley b. 1945

The COBE Satellite



DIRBE Solar Elongation 90° Maps: Mid-Infrared



DMR's Two Year CMB Anisotropy Result



Center for Astrophysics (CfA) survey

10,506 galaxies in the cone-shaped survey region, which extends out to 135 megaparsecs in the northern hemisphere, with the earth at the apex of the cone.



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FIG. 1.-50% high volume contours from three galaxy surveys across three decades. From left to right, they are Gott, Melott, & Dickinson (1986), Vogeley et al. (1994), and the present work.



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Mapping the brain



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A general formulation

Excursion sets



$A_u \equiv A_u(f, M) \stackrel{\Delta}{=} \{t \in M : f(t) \geq u\}$

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A more general setting



$A_D \equiv A_D(f, M) \stackrel{\Delta}{=} \{t \in M : f(t) \in D\}$

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Geometry



$$\lambda_2 (\text{Tube}(M, \rho)) = \pi \rho^2 + \rho \times 4L + L^2$$

$$= \sum_{j=0}^{2} \omega_{2-j} \rho^{2-j} \mathcal{L}_j(M)$$

where

$$\omega_j = \frac{\pi^{j/2}}{\Gamma(\frac{j}{2}+1)} = \frac{s_j}{j}$$



 λ_3 (Tube(M, ρ))

$$= \frac{4}{3}\pi\rho^{3} + 12 \cdot \frac{1}{4}\pi\rho^{2} \cdot L + 6\rho L^{2} + L^{3}$$
$$= \sum_{j=0}^{3} \omega_{3-j}\rho^{3-j}\mathcal{L}_{j}(M)$$

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$$\lambda_3 (\operatorname{Tube}(M, \rho)) = \sum_{j=0}^2 \omega_{3-j} \rho^{3-j} \mathcal{L}_j(M)$$

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Steiner's Formula



1796-1863, Switzerland

For nice (e.g. convex) $M \in \mathbb{R}^N$, and $N' \ge N$, the volume of $\operatorname{Tube}(M, \rho) = \left\{ t \in \mathbb{R}^{N'} : d_{N'}(t, M) \le \rho \right\}$

is, for $\rho < \rho_c(M)$, given by,

$$\lambda_{N'} (\operatorname{Tube}(M, \rho)) = \sum_{j=0}^{N} \omega_{N'-j} \rho^{N'-j} \mathcal{L}_j(M)$$

The \mathcal{L}_i can be defined via the tube formula and are *intrinsic*.

$$\lambda_{N'} (\operatorname{Tube}(M, \rho)) = \sum_{j=0}^{N} \omega_{N'-j} \rho^{N'-j} \mathcal{L}_j(M)$$



Wilhelm Killing Germany 1847-1923



Rudolf Lipschitz Germany 1832-1903



Hermann Weyl Germany/USA 1885-1951

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Treat M as a Riemannian manifold

Curvature tensor:,

$$R(X, Y, Z, W) = (\nabla_X \nabla_Y Z - \nabla_Y \nabla_X Z - \nabla_{[X,Y]} Z, W)$$

Second fundamental form S

$$S(X,Y) \stackrel{\Delta}{=} \widehat{\nabla}_X Y - \nabla_X Y = P_{TM}^{\perp} \left(\widehat{\nabla}_X Y \right)$$

Scalar second fundamental form $S_{
u}$

$$S_{\nu}(X,Y) \stackrel{\Delta}{=} (S(X,Y),\nu)$$

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(For ν a unit normal vector field on M)

LKC's: The general case

$$\mathcal{L}_{i}(M) = \sum_{j=i}^{N} \sum_{m=0}^{\lfloor \frac{j-i}{2} \rfloor} C_{Nnim} \int_{\partial_{j}M} \int_{S(T_{t}\partial_{j}M^{\perp})} \operatorname{Tr}^{T_{t}\partial_{j}M} \left(\widehat{R}^{m}\widehat{S}_{\nu_{N-j}}^{j-i-2m}\right) \\ \times 1_{N_{t}M}(-\nu_{N-j}) \mathcal{H}_{N-j-1}(d\nu_{N-j}) \mathcal{H}_{j}(dt)$$

LKC's: The general case

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Leonhard Euler Switzerland 1707-1783



Jules Henri Poincaré France 1854-1912

\mathcal{L}_0 : The Euler-Poincaré characteristic

 $M \subset \mathbb{R}^N$ is nice, of dimension k, and "triangulisable" $\alpha_0 =$ number of vertices $\alpha_1 =$ number of lines

 $\alpha_k =$ number of "full" simplices in the triangulation

 $\mathcal{L}_0(M) \equiv$ Euler characteristic of M is

$$\varphi(M) = \alpha_0 - \alpha_1 + \dots + (-1)^d \alpha_N$$

3-d excursion sets

Meatball, EC=21



Sponge, EC=-15



Bubble, EC=1



Averaged geometry of

excursion sets

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Excursion sets



$A_u \equiv A_u(f, M) \stackrel{\Delta}{=} \{t \in M : f(t) \geq u\}$

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A 30-year old formula

Suppose f is Gaussian, mean zero, variance σ^2 , stationary, and isotropic, with second spectral moment λ_2 and $M = [0, T]^N$ Then:

$$\mathbb{E}\left\{\mathcal{L}_0\left(A_u\right)\right\} = e^{-u^2/2\sigma^2} \sum_{k=1}^N \frac{\binom{N}{k} T^k \lambda_2^{k/2}}{(2\pi)^{(k+1)/2} \sigma^k} H_{k-1}\left(\frac{u}{\sigma}\right) + \Psi\left(\frac{u}{\sigma}\right).$$

where

$$H_n(x) = n! \sum_{j=0}^{\lfloor n/2 \rfloor} \frac{(-1)^j x^{n-2j}}{j! (n-2j)! 2^j}, \qquad n \ge 0, \ x \in \mathbb{R}$$
$$\Psi(x) = \frac{1}{\sqrt{2\pi}} \int_x^\infty e^{-x^2/2} dx$$

One dimension: A line of length T

$$\mathbb{E} \{ \mathcal{L}_0 (A_u(f, [0, T])) \} = \Psi(u/\sigma) + \frac{T \lambda_2^{1/2}}{2\pi\sigma} e^{-u^2/2\sigma^2},$$



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Two dimensions: A square of side length T



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Three dimensions: A cube of side length T

$$\left[\frac{T^{3}\lambda_{2}^{3/2}}{(2\pi)^{2}}u^{2} + \frac{3T^{2}\lambda_{2}}{(2\pi)^{3/2}}u + \frac{3T\lambda_{2}^{1/2}}{2\pi} - \frac{T^{3}\lambda_{2}^{3/2}}{(2\pi)^{2}}\right]e^{-u^{2}/2} + \Psi(u).$$


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The observed EC of the set of high-density regions of the CfA Galaxy survey. Also shown is the expected EC for randomly distributed galaxies with no structure; the CfA data has smaller EC than expected, indicating less "blobs" and more clumping of galaxies into clusters, strings, and "walls".



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FIG. 7.— Genus curves with shaded 1σ error regions for the (a) 100 DH and (b) 50 MR samples, compared with SDSS and Gaussian ndom phase.

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DMR's Two Year CMB Anisotropy Result





threshold, t

Figure 12. Plot of the observed EC of excursion sets of the anomalies in the cosmic microwave background radiation (jagged line), and the expected EC from the formula (smooth line) if there are no real anomalies. The observed microwave background radiation produces an EC curve similar in shape to that expected, but somewhat lower and spread more in the tails—evidence that some of the anomalies are real and not just due to random noise. This discrepancy points to a Gaussian random field model for the anomalies, with a larger standard deviation and a larger smoothness than the background noise.



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Excursion probabilities

$$\mathbb{P}\left\{\sup_{t\in M}f(t)\geq u\right\}$$

$$\sim \mathbb{E}\left\{\mathcal{L}_0\left(A_u(f,M)\right)\right\}$$

$$\liminf_{u\to\infty} u^{-2}\log|\mathbb{P}-\mathbb{E}| \geq \frac{1}{2} + \frac{1}{2\sigma^2(f)}$$

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Observed and expected EC for the PET data and the expected EC if there is no activation due to the linguistic task. In particular, at u = 3.3 we expect an EC of 1, but we observe 4. At the 5% critical value of u = 4.22, we expect 0.05 but we observe 2 components.

A more general result



$$A_D \equiv A_D(f, M) \stackrel{\Delta}{=} \{t \in M : f(t) \in D\}$$

$$\mathbb{E}\left\{\mathcal{L}_{j}(A_{D})\right\} = \sum_{I=0}^{N-j} \begin{bmatrix} j+I\\I \end{bmatrix} (2\pi)^{-j/2} \mathcal{L}_{j+I}(M) \mathcal{M}_{I}^{(k)}(D)$$

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Non-Gaussian fields



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Now for the mathematics

A more general result



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Gaussian tube formula Gauss measure on \mathbb{R}^k

$$\gamma_k(D) \stackrel{\Delta}{=} \frac{1}{(2\pi)^{k/2}} \int_D e^{-\|x\|^2/2} dx$$

Gaussian tube formula



$$\gamma_k(\operatorname{Tube}(D,\rho)) = \gamma_k(D) + \sum_{j=1}^{\infty} \frac{\rho^j}{j!} \mathcal{M}_j^{\gamma}(D)$$

 $\mathbb{E}\left\{\mathcal{L}_{j}(A_{D})\right\} = \sum_{l=0}^{N-j} \begin{bmatrix} j+l\\l \end{bmatrix} (2\pi)^{-j/2} \mathcal{L}_{j+l}(M) \mathcal{M}_{l}^{(k)}(D)$

Turn M into a Riemannian manifold with

$$g_t(X_t, Y_t) = \mathbb{E} \{X_t f_t \cdot Y_t f_t\}$$



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Scalar second fundamental form S_{ν}

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(For ν a unit normal vector field on M)

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u_{N-j})\mathcal{H}_j(dt)$

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Kinematic fundamental formula



Kinematic fundamental formula







Kinematic fundamental formula





$$\int \mathcal{L}_i \left(M_1 \cap g M_2 \right) \, d\nu(g) = \sum_{j=0}^{N-i} \begin{bmatrix} i+j\\i \end{bmatrix} \begin{bmatrix} N\\j \end{bmatrix}^{-1} \mathcal{L}_{i+j}(M_1) \mathcal{L}_{N-j}(M_2)$$

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A more general result



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Crofton's formula on \mathbb{R}^N



Crofton's formula on \mathbb{R}^N



$$\int_{\mathrm{Graff}(N,N-k)} \mathcal{L}_j(M \cap V) \ d\lambda_{N-k}^N(V) = \begin{bmatrix} k+j \\ j \end{bmatrix} \mathcal{L}_{k+j}(M)$$

A Gauss-Crofton formula

M a C^2 , *n*-dimensional, Riemannian manifold y^1, \ldots, y^k Gaussian on M, matched to the metric Define, for $u \in \mathbb{R}^k$, the (random) submanifold

$$D_u = \left\{ t \in M : y_t^1 = u_1, \dots, y_t^k = u_k \right\}$$

Take $Z_k \in \mathbb{R}^k$ standard Gaussian independent of y

$$D_{Z_k} = \{t \in M : y_t = Z_k\}$$

$$\mathbb{E} \{ \mathcal{L}_{j}(M \cap D_{Z_{k}}) \} = (2\pi)^{-k/2} \frac{[k+j]!}{[j]!} \mathcal{L}_{k+j}(M)$$

About the proofs

Excursion probabilities

$$\mathbb{P}\left\{\sup_{t\in M}f(t)\geq u\right\}$$

$$\sim \mathbb{E}\left\{\mathcal{L}_0\left(A_u(f,M)\right)\right\}$$

$$\liminf_{u\to\infty} u^{-2}\log|\mathbb{P}-\mathbb{E}| \geq \frac{1}{2} + \frac{1}{2\sigma^2(f)}$$

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$$f_t = \sum_{n=1}^{K} \xi_n \varphi_n(t) = \langle \xi, \varphi(t) \rangle$$

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$$f_t = \sum_{n=1}^{K} \xi_n \varphi_n(t) = \langle \xi, \varphi(t) \rangle$$
$$1 = \mathbb{E} \{ f_t^2 \} = \sum_{1}^{K} \varphi_j^2(t) = \| \varphi(t) \|$$

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$$f_{t} = \sum_{n=1}^{K} \xi_{n} \varphi_{n}(t) = \langle \xi, \varphi(t) \rangle$$
$$1 = \mathbb{E} \{ f_{t}^{2} \} = \sum_{1}^{K} \varphi_{j}^{2}(t) = \| \varphi(t) \|$$

Map M into the sphere with

$$t \rightarrow \varphi(t)$$

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$$f_{t} = \sum_{n=1}^{K} \xi_{n} \varphi_{n}(t) = \langle \xi, \varphi(t) \rangle$$
$$1 = \mathbb{E} \{ f_{t}^{2} \} = \sum_{1}^{K} \varphi_{j}^{2}(t) = ||\varphi(t)|$$

Map M into the sphere with

$$t \rightarrow \varphi(t)$$

Then

$$g(x) \stackrel{\Delta}{=} f(\varphi^{-1}(x))$$

has covariance

 $\mathbb{E}\left\{g(x)g(y)\right\} = \langle x, y \rangle$

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One (Gaussian) case covers all !?!



 $t \rightarrow \varphi(t) \stackrel{\Delta}{=} (\varphi_1(t), \dots, \varphi_K(t))$

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Laboratoire de MATHEMATIQUES



Bibliothèque

Plan d'accés

Menu





Photographe : J.F. FERRATON - Saint-Flour

Founded in 1971, this school is organised every year by the Laboratoire de Mathématiques (UMR 6620). It is supported by Blaise Pascal University (Clemont-Ferrand II), the Ministry of Research and the C.N.R.S. It is intended for PhD students, teachers and researchers who are interested in probability theory, statistics, and in applications of these techniques.

The school has three main goals:

- 1. to provide, in three high level courses, a comprehensive study of a field in probability theory or statistics;
- 2 to enable the participants to evolain their work in lectures:



Youcef Amirat Tél : +33 (0)4 73 40 70 62





Valérie Sourlier Tél : +33 (0)4 73 40 **70 50** Fax : +33 (0)4 73 40 54 50

Informatique

Damien Ferney Tél : +33 (0)4 73 40 70 68

Cédric Barrel Tél : +33 (0)4 73 40 70 55

Adresse

Laboratoire de Mathématiques Université Blaise Pascal Campus Universitaire des Cézeaux 63177 Aubière cedex France

Fax : +33 (0)4 73 40 70 64

Webmaster

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Lectures 2009

Robert ADLER: Topological complexity of smooth random functions. Mireille BOUSQUET-MÉLOU: Enumerative combinatorics for probability. Alison ETHERIDGE: Some mathematical models from population genetics.

Abstracts

Practical information and registration rate

The participants will be lodged at the Maison des Planchettes, 7 rue des Planchettes, 15100 Saint-Flour (France). The lectures will be given at the same place. Full board accommodation is included in the registration fee for the participants; the families can also be lodged, and their accommodation should be paid for during the school.

More information, in particular the registration fee, will be available in February.

Registration

SMM

ADLER

Random Fields and Geometry

ROBERT J. ADLER JONATHAN E. TAYLOR

This monograph is devoted to a completely new approach to geometric problems arising in the study of random fields. The groundbreaking material in Part II, for which the background is carefully prepared in Parts I and II, is of both theoretical and practical importance, and stirking in the way in which problems arising in geometry and probability are beautifully intertwined.

The three parts to the monograph are quite distinct. Part I presents a user-friendly yet comprehensive background to the general theory of Gaussian random fields, treating classical topics such as continuity and boundedness, entropy and majorizing measures, horall and Skeptin integuilate. Part I gives a quick review of geometry, both integral and Riemannian, to provide the reader with the material meeded for Part III, and to give some new results and new proofs of known results along the way. Topics such as Conton formulae, curvature measures for stratified manifolds, critical point formulae, curvature measures for stratified manifolds, relical point concise, self-contained treatment of all of the above topics, which are necessary for the study of random fields. The new approach in Part II is devoted to the geometry of eccarion sets of random fields and the related failer characteristic approach to extremal probabilities.

Random Fields and Geometry will be useful for probabilities and statisticians, and for theoretical and applied mathematicians who wish to learn about new relationships between geometry and probability. It will be helpful for graduate students in a classroom setting, or for self-study. Finally, this text will serve as a basic reference for all those interested in the companion volume of the applications of the theory. These applications, to appear in a forthcoming volume, will corer areas as widespread as brain imaging, physical ocennography, and astrophysics.



Random Fields and Geometry





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