**Geometrical, Algebraic, Functional and Correlation Inequalities Applied in Support of James-Stein Estimator for Multidimensional Projections**

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**Abstract:** Isoperimetric, Milman reverse, Hilbert, Widder, Fan-Taussky-Todd, Landau, and Fortuin–Kasteleyn–Ginibre (FKG) inequalities in n dimensions in investigations of multidimensional estimators support the use of James-Stein estimator against classical least squares as applied to Cumulant Analysis, Associate Random Variables, and Time Series Analysis.

**Keywords:** Multidimensional Time Model, James-Stein Estimator, Sampling and Functional Inequalities

1. Introduction

“quoting Virgil:
At last they landed, where from far your eyes
May view the turrets of new Carthage rise;
There bought a space of ground, which Byrsa call’d,
From the bull’s hide they first inclos’d, and wall’d.

*(Aeneid, Dryden’s translation)*

This refers to the legend of Dido. Virgil’s version has it that Dido, daughter of the king of Tyre, fled her home after her brother had killed her husband. Then she ended up on the north coast of Africa, where she bargained to buy as much land as she could enclose with an oxhide. So she cut the hide into thin strips, and then she faced, and presumably solved, the problem of enclosing the largest possible area within a given perimeter—the isoperimetric problem. But earthly factors mar the purity of the problem, for surely the clever Dido would have chosen an area by the coast so as to exploit the shore as part of the perimeter. This is essential for the mathematics as well as for the progress of the story. Virgil tells us that Aeneas, on his quest to found Rome, is shipwrecked and blown ashore at Carthage. Dido falls in love with him, but he does not return her love. He sails away and Dido kills herself. Kline concludes [23, p. 135]:

“And so an ungrateful and un receptive man with a rigid mind caused the loss of a potential mathematician. This was the first blow to mathematics which the Romans dealt.” [33]

From this last quote of Kline we can think or even deduce without rigorous proof that Dido possibly suspected some mistake or maybe something mysteriously unsolved in the problem, which she could blame for her unsuccessful love.

The subject of multidimensional projections that is equivalent to reduction of dimensions has a long history and there are several difficulties of approaching it. The development of multidimensional mathematical and statistical apparatus should not be ignored, when considering reduction in number of dimensions in simplifying mathematical or statistical problem. The very good example are James-Stein estimators that depend on the number of dimensions and in dimensions more than 2 is deferent from the classical least squares estimator that almost uniformly reduces every problem to 2-dimensional (2D) consideration.

2. Inequalities as the Method of Study the Above Mentioned Problem

This article offers as a method of the direct approach to the above mentioned problem consideration of Geometrical, Algebraic, Functional and Correlation Inequalities in 2 and more dimensions and various approaches for their proofs.
2.1. Isoperimetric Inequalities

The Isoperimetric inequality for convex body K in $\mathbb{R}^n$, $B^n$ n-dimensional ball of unit radius and S a surface area of a compact set in $\mathbb{R}^n$:

$$\left( \frac{\text{Vol}(K)}{\text{Vol}(B^n)} \right)^{1/n} \leq \left( \frac{S(K)}{S(B^n)} \right)^{1/(n-1)}$$

(1)

Where Volume and Surface area of the ball of radius r in $\mathbb{R}^n$ is

$$\text{Vol}(B^n(r)) = \frac{\pi^{n/2} r^n}{\Gamma(\frac{n}{2}+1)}$$

and $S(B^n(r)) = \frac{2\pi^{n/2} r^{n-1}}{\Gamma(\frac{n}{2})}$

(2)

The solution of problem is traced back to Pythagoras (2600 - 2500 years ago), who first approached the problem, and then to Aristotle’s (200 years later) knowledge of the maximum principle of the circle. Zenodorus (200–140 B.C.) (though his monograph was lost) was the first to write the ternary function defined on the interval $[0; 1]$ are known to already been shown by Lhuilier using differential calculus.

According to [33] maximization of the area of two with fixed perimeter has the greatest area, and if the 2 isosceles triangles are not counterexample to Zenodorus lemma the isosceles triangle and the bases $a, b$ can lead to contradiction argument.

For example, the well-known Volterra function and Cantor ternary function defined on the interval $[0; 1]$ are known to be continuous and rectifiable, but the arc length integral does not exist either as a proper or improper Riemann integral. It uses a Theorem if $y$ and $y′$ belong to the Euler class, then so do the variations $y + t y′$ [3]. Gergonne some 200 years ago used symmetry argument that was later used by Steiner. The first formal proof for the planar case is attributed to Steiner (170 years ago) who gave five different proofs [33-37].

It can easily be derived from Brunn-Minkowski inequality, which however, has its inverse inequality found by Milman [48, 49].

For non rigorous consideration the proof for plane case that $A \leq L^2/4\pi$ consists from few steps, where A denotes the area, and L the length of the curve.

1. Lemma. The figure solving the isoperimetric problem is convex. The proof uses contradiction argument.

2. Consideration of the n-polygons that consist from triangles.

3. Theorem. Only isosceles triangles should be considered.

4. Rescaling for inscribing the cycle into the n-polygon.

5. According to [33] maximization of the area of two isosceles triangles with given bases and the base angles $\alpha, \beta$ and the bases $a, b$: $\sin \alpha: \sin \beta = a: b$ can lead to counterexample to Zenodorus lemma the isosceles triangle has the greatest area, and if the 2 isosceles triangles are not similar the similar isosceles triangles with the same bases and with the same sum of perimeters have the greatest area.

That was shown geometrically by Steiner, who says it had already been shown by Lhuilier using differential calculus.

6. In contrast to the earlier arguments by symmetry Steiner’s argument is based on mean curve that is equally distanced from 2 original curves.

7. Another approach considers maximizing the area for curves $t \rightarrow (x (t), y (t))$ (that look like the method of proof of Han-Banach Theorem)

$$\int A(x, x', y, y', t) \, dt = 1/2 \int (xy' - x'y)\,dt$$

(3)

With perimeter

$$\int L(x, x', y, y', t) \, dt = \int \sqrt{x'^2 + y'^2}\,dt$$

(4)

using Lagrange multipliers and solving Eulers equations.

8. Another approach for n-polygons the problem becomes to maximize

$$A = 1/2 \sum_{k=1}^{n-1} x_k y_{k+1} - y_k x_{k+1}$$

(5)

with fixed perimeter

$$L = \sum_{k=1}^{n-1} \sqrt{(x_k - x_{k-1})^2 + (y_k - y_{k-1})^2}$$

(6)

9. A year after Charles Hermite death Hurwitz [1] published Fourier series proof

$$f (t) = 1/2a_0 + \sum_{n=1} a_n \cos (nt) + b_n \sin (nt)$$

(7)

With application of Parseval’s theorem

$$\int_0^{2\pi} f(x) g(x) \, dx = a_0 b_0 /2 + \sum_{n=1} a_n b_n + c_n d_n$$

(8)

The problem with the above parameterization $t \rightarrow (x (t), y (t))$ becomes

$$x'^2 + y'^2 = (L/2\pi)^2$$

(9)

with $x, x', y,$ and $y'$ as below

$$x (t) = 1/2a_0 + \sum_{n=1} a_n \cos (nt) + b_n \sin (nt)$$

(10)

$$y (t) = 1/2c_0 + \sum_{n=1} c_n \cos (nt) + d_n \sin (nt)$$

(11)

$$x'(t) = \sum_{n=1} b_n \cos (nt) - a_n \sin (nt)$$

(12)

$$y'(t) = \sum_{n=1} c_n \cos (nt) - c_n \sin (nt)$$

(13)

After expansion for

$$L^2 = 4\pi A$$

(14)

term for $n=1$ becomes $a_1 = d_1$ and $b_1 = -c_1$ with terms for $n > 1$ vanishing $a_n = d_n = b_n = c_n = 0$ and the equations become equations for circle.

10. After 20 years Carleman published proof based on power series theorem

$$I = \int_0^{2\pi} f(\varphi(z)) \, d\theta$$

and

$$A = \int_0^{2\pi} \int_0^L | \varphi'(z)|^2 \, d\rho \, d\theta$$

(15)

Consider analytic function in unit disk $\psi \equiv \varphi' \varphi$ (z)

$$\psi \equiv \varphi' \varphi (z) = \sum_{n=0} a_n z^n$$

Let $b_n = a_0 a_n + a_1 a_{n+1} + \ldots a_n a_1$, and $(\sum_n a_n z^n) = \sum_n b_n z^n$

Then $L = 2\pi \sum |a_n|^2$ with

$$A = \int_0^{2\pi} \int_0^L |b_n|^2 r^2 \, r \, dr \, d\theta = \pi \sum |b_n|^2 /n+1$$

And the isoperimetric inequality becomes

$$\sum_{n=0} |b_n|^2 /n+1 \leq (\sum_{n=0} |a_n|^2) (\sum_{n=0} |b_n|^2)$$

(16)
It was mentioned before that Isoperimetric inequality can be derived from Brunn–Minkowski inequality that was proposed by Hermann Brunn 10 years before Minkowski and states:

$$\mu(A + B)^{1/n} \geq (\mu(A))^{1/n} + (\mu(B))^{1/n},$$ (17)

where $\mu$ denotes $n$-dimensional Lebesgue measure and the $+$ on the left-hand side denotes Minkowski addition.

99 years passed before V. Milman established reverse form

$$\text{Vol}(s \varphi K + t \psi L)^{1/n} \leq C \text{Vol}(s \varphi K)^{1/n} + t \text{Vol}(\psi L)^{1/n},$$ (18)

where $\varphi$ and $\psi$ are volume-preserving linear maps from $\mathbb{R}^n$ to itself for any real numbers $s, t > 0$.

### 2.2. Sampling and Functional Inequalities Related to Probability and Stochastic Inequalities

Some of the most interesting extensions were inequalities of Cauchy and Hilbert. Hilbert some 110 years ago, though did not use Cauchy’s inequality, but rather some complicated methods to prove that

$$\sum a_m b_n \leq (\sum a_m^2)^{1/2} (\sum b_n^2)^{1/2},$$ (19)

with $C = 2\pi$, where the summations run from 1 to $+\infty$, which was later reduced by I. Schur to $C = \pi$.

This inequality inspired very famous mathematicians G. H. Hardy, J. E. Littlewood and G. Pólya so much, that not only they devoted 2 (two) last chapters of their book “Inequalities” to Hilbert’s double series theorem and its seemingly statistical extension “Rearrangements”, but also as it was discussed in “by Hardy in his “Prolegomena to a chapter on inequalities,” which was a Presidential Address at the annual meeting of the London Mathematical Society of 8 November/28, 1 year before D. V. Widder introduced two

$$\sum_{k=1}^{n}(a_1 a_2 \ldots a_k)^{1/k} \leq C \sum_{k=1}^{n} a_k,$$

$$C_n = 4 - 12 \pi^2/(n^2) + O (\ln \ln n/n^2),$$ (20)

And for the Hardy’s inequality

$$C_n = 4 - 16 \pi^2 / (n^2) + O (\ln \ln n/n^3),$$ (21)

It is interesting to notice that Wilf advised Bruijn to apply recurrence argument to Carleman’s method.

The extensions of Widder with the remark of Hardy are the following inequalities:

$$\sum \frac{\log(m/n)}{m-n} a_m b_n \leq \pi \left( \sum \frac{a_m b_n}{m+n} \right)^{1/2} \left( \sum \frac{b_m b_n}{m+n} \right)^{1/2},$$ (25)

$$\sum \frac{\log(m/n)}{\max(m,n)} a_m b_n \leq 2 \left( \sum \frac{a_m b_n}{\max(m,n)} \right)^{1/2} \left( \sum \frac{b_m b_n}{\max(m,n)} \right)^{1/2},$$ (26)

where the summations run from 0 to $+\infty$, and the coefficient on the left-hand side is interpreted as $1/n$ when $m = n$.

In this form it is very much resembling the famous Landau-Kolmogorov inequality, which was introduced by E. Landau 2 years after I. Schur’s improvement.

$$\|f''\| \leq 4 \|f\| \|f''\|$$ (27)

He showed that for the norm of $\|f\|$ defined to be the supremum of $|f(x)|$, a real-valued function $f$ defined on $(0, \infty)$. If $f$ is twice differentiable and both $f$ and $f''$ are bounded:

$$\|f''\| \leq 2 \|f\|^{1/2} \|f''\|^{1/2}$$ (28)

It was extended by A. N. Kolmogorov 5 years after the book by G. H. Hardy, J. E. Littlewood and G. Pólya, on inequalities and 10 years later was translated in Math Ann.
For determination of the best constants $C(n, k)$ for the $n$-th
derivative of $f$ defined on $(0, \infty)$ if both $f$ and $f^{(n)}$ are
bounded.

\[ \|f^{(k)}\| \leq C(n, k) \|f\|^{1-k/n} \|f^{(n)}\|^{k/n}, \]
where $C(n, k) = a_{n+k} a_n^{-1+k/n}$

2.3. Introduction of Stein’s Estimators for Exponential
Family [46-48]

Introduction of Stein’s estimators for exponential family of
distributions is another connection between Poisson process in
dimension $\geq 3$ and normal or Gaussian distributions. If $X$
is from exponential family of distributions its density is given
by functional dependence:

\[ f_\theta(x) = \exp\{\theta x - \phi(\theta)\} \kappa(x), \ x \in \mathbb{R}, \]

let $t(X) = -k'(\theta)/k(\theta)$ for any absolutely continuous function $g$
on $\mathbb{R}$, such that $E[g'(X)] < \infty$, then the following identity holds

\[ E\{t(X) - \theta\}g(X) = E\{g'(X)\}. \]

This important identity is very useful for the application of
the methods of linear combinations and ratios of random
variables of exponential family of distributions that would be
discussed in Subsection 2.3.1. and for future development of
the theory of Stein estimator and different tests of hypotheses
for multidimensional exponential families.

2.4. Linear Regression Model and Stein’s Estimator
Compared to Least Squares Estimator for
Multidimensional Exponential Families

For regression model, $Y = XC + e,$

\[ Y = XC + e, \]

where $Y$ is a $N \times 1$ vector of $N$ observations on the variable
to be explained, $X$ is a $N \times K$ full-column-rank matrix of $N$
observations on $K$ fixed explanatory variables, $C$ is a column
vector of regression coefficients, and $e$ is a $N \times 1$ error vector
with a multivariate normal distribution with mean vector 0
and variance-covariance matrix $\Sigma_{xx}$, with $\Sigma$ unknown. The
least-squares (LS) estimator for $C$
is

\[ \hat{C}_{LS} = (X'X)^{-1} X Y; \]

Stein’s estimator compared to Least Squares estimator for
multidimensional exponential families

Stein’s estimator for the above regression model is:

\[ C_{SE} = \left[1 - \frac{(Y' - X'C') (Y' - X'C)}{C'X'C} \right] C, \]

where $L$ is a scalar,

\[ n = N-K, \quad S^2 = (Y' - X'C)' (Y' - X'C)/n \]

Though $C_{SE}$ is also biased, it dominates $\hat{C}_{LS}$ so as

\[ E\{(C_{SE} - C)' X'X (C_{SE} - C)\} < E\{(\hat{C}_{LS} - C)' X'X (\hat{C}_{LS} - C)\}, \quad 0 < L < \frac{(K-2)}{(n+2)}, K > 2. \]

An unbiased estimator of the bias of

\[ C_{SE} \text{ is } C_{UB} (C_{SE}) = C_{SE} - \hat{C}_{LS} = -L \frac{nS^2}{C_{LS}'X'C_{LS}} \hat{C}_{LS} \]

(42)
2.5. Properties of Statistics and the Uniformly Most Powerful Invariant Test

It is worth to mention property of independence of

\[ Z = (X'X)^{1/2} \tilde{C}_{LS} \propto N(\mu, \sigma^2 I_k) \quad (\mu = (X'X)^{1/2}C) \text{and} \quad (nS^2/\sigma^2) \propto \chi^2_n. \]

To test null hypothesis \( H_0: C = 0 \) against alternative hypothesis \( H_1: C \neq 0 \), the uniformly most powerful invariant test statistic is

\[ F = (\tilde{C}_{LS}'X'X\tilde{C}_{LS})/KS^2, \quad F \propto F_{K,n} \]

an F-distribution with \( K \) and \( n \) degrees of freedom and noncentrality parameter, \( D = C'X'C/2\sigma^2 \).

And for F-ratio based on \( C_{SE} \) statistic

\[ F = C_{SE}\left[ (S^2(XX')^{-1})C_{SE} \right] = F \left( 1 - nL/KF \right)^2 \]

\( F \) is a function of \( F \), and it is invariant to the same linear and orthogonal transformations, as \( F \) is.

3. Simple Rules for the Reduction of Dimensions

As a result of the above discussion that established the support for the James – Stein estimators from Geometrical, Algebraic, Functional and Correlation Inequalities the following rules can be used for the consideration in the problems involving the reduction of dimensions:

1. The multidimensional problem should not be reduced to less than 3-dimensional problem;
2. James – Stein estimators should be preferred in place of usual least Squares estimators;
3. Other types of approaches to model multidimensional problem should be explored.

As an example below is a supporting discussion on three approaches to Multidimensional Time Model for Probability Distribution Function (MTM for PDF). It is concluded by Proposition that allows a new perspective in approaching multidimensional problem as decomposition of a process with special properties into composition of processes of Brownian motion.

4. First Approach to Multidimensional Time Model Through Kramers Turnover Problem in the Theory of Velocity of Chemical Reactions

Consider first the mathematical structure of the models of Boltzmann type kinetic equations for reacting gas mixtures for particles undergoing inelastic interactions with reactions of bimolecular and dissociation-recombination type is very complicated, because of the collisional operators that usually in the full Boltzmann equations, are expressed by 5-fold integrals. Consequently direct numerical applications of these models present several computational difficulties. The search for the simpler solution had its long way till the introduction of the equation for the Brownian motion by Albert Einstein. However, using the theory of Brownian motion for the velocity (rate) of chemical reactions Bohr, Kramers, and Slater used only one-dimensional (1D) model for the Kramers turnover problem, that is, obtaining a uniform expression for the rate of escape of a particle over a barrier for any value of the external friction until it was corrected by Grote-Hynes theory 40 years later, with new improvements following after 6 years by Mel'nikov and Meshkov (MM). There are certainly other theories followed, all of them distinguish 1D approach from 2D, 3D, and multiD approaches.

It is important and very interesting to consider such point that Kramers in his original work had it as possibility that multidimensional pattern could be related to time dimensions, as he based his introduction theory of Brownian motion on the Einstein’s pattern he considered a range of time intervals \( \tau \). His discussion of the possibility of a term proportional to \( \tau \) in the expression for Moments of Brownian motion \( B_{\tau n}(n>1) \) related it to the fact that “the values, which \( X \) takes at moments \( t_1, t_2, \ldots, t_n \) which lie sufficiently close together are no longer independent; and Moments of Brownian motion \( B_{\tau n}(n>1) \) in fact are represented by a volume integral

\[ \int \ldots \int X(t_1)X(t_2)\ldots X(t_n)dt_1dt_2\ldots dt_n \]

over an \( n \)-dimensional cube; the contribution to this integral due to a narrow cylinder extending along the diagonal \( t_1=t_2=\ldots=t_n \) may give a term proportional to \( \tau \) ” [11]

4.1. Second Approach to Multidimensional Time Model Through Cumulant Functions and Time Series Analysis (Brillinger)

To strengthen this notion consider cumulants properties for time series analysis that provide measure of Gaussianity. If \( r \) v. \( X \) is normal, then \( \text{cum}_k \{X\} = 0 \) for \( k > 2 \), where \( \text{cum}_k \) denotes the joint cumulants of \( X \) with itself \( k \) times.

For simplicity consider seq of iid \( X_i \) with all moments and \( E \{X_i\} = 0 \) and var \( \{X_i\} = 1 \), then for \( S_n = \sum X_i/N \), \( \text{cum}_k \{S_n\} = n^{k/2} \text{cum}_k \{X_i\}/n^{k/2} \) that tends to 0 for \( k > 2 \), as \( n \) tends to infinity, so \( S_n \) has a limiting normal distribution.

And for time series analysis the moment function \( E \{X(t+u)\} \ldots X(t+u(k-1))X(t)\} \) would not depend on \( t \), and on short time interval centered at point of time \( t \) can be approximated by normal distribution.

4.2. Third Approach Through Associated Random Variables [23, 24]

Additional to the Brownian motion considerations in the theory of chemical reactions and time series analysis for
cumulant functions, the same results can be obtained from the consideration of associated random variables.

[49] Consideration of associated random variables can be supported by Fortuin–Kasteleyn–Ginibre (FKG)

\[ \left( \sum_{x \in X} f(x) g(x) \mu(x) \right) \geq \left( \sum_{x \in X} \mu(x) \right) \]

Or \( f_g - f_e \geq 0 \), where finite distributive lattice has a least element, a minimal element \( x \neq 0 \) is called an atom and relations \( \Lambda \) and \( \Xi \) satisfy either of the following

\[ x \Lambda (y \lor z) = (x \Lambda y) \lor (x \Lambda z) \]

for all \( x, y, z \) in \( X \)

Definition 1. For \( n > 1 \) the set of rv \( X \) is said to be associated, if for all given real-valued functions \( g_i \) that are increasing ineach component when the other components are held fixed, the inequality

\[ E \left[ \prod_{i=1}^n f_i(X) \right] \geq E \left[ \prod_{i=1}^n g_i(X) \right] \text{holds, or equivalently,} \]

\[ \text{Corr} \left( g_i(X), g_j(X) \right) \geq 0, \]

Theorem 1. (a) A set consisting of a single random variable is a set of associated random variables. (b) Independent random variables are associated random variables. (c) A subset of a set of associated random variables forms a set of associated random variables. (d) Increasing functions of associated random variables are associated random variables [24].

Proposition 2. Therefore, the process \( X(t) \) with above properties can be represented by composition of Brownian motion processes in finite-dimensional time model.

\[
\mu(x \land y) \geq \mu(x) \land \mu(y) \quad \forall \ x, y \in \mathbb{X}
\]

Inequality as a correlation inequality that states for finite distributive lattice on \( \mathbb{X} \) with \( \mu \) nonnegative function on it satisfying FKG lattice conditions

\[ \mu(x \land y) \geq \mu(x) \land \mu(y) \quad \forall \ x, y \in \mathbb{X} \]

References

[1] M. Fundator Applications of Multidimensional Time Model for Probability Cumulative Function for Parameter and Risk Reduction. In JSM Proceedings Health Policy Statistics Section Alexandria, VA: American Statistical Association. 433-441.

[2] M. Fundator. Multidimensional Time Model for Probability Cumulative Function. In JSM Proceedings Health Policy Statistics Section. 4029-4039.

[3] M. Fundator. Testing Statistical Hypothesis in Light of Mathematical Aspects in Analysis of Probability doi:10.20944/preprints201607.0069.v1

[4] M. Fundator Application of Multidimensional time model for probability Cumulative Function to Brownian motion on fractals in chemical reactions (44th Middle Atlantic Regional Meeting, June/9-12/16, Riverdale, NY) Academia Journal of Scientific Research (ISSN 2315-7712) DOI: 10.15413/ajsr.2016.0167 In preparation for publication.

[5] Michael Fundator Application of Multidimensional time model for probability Cumulative Function to Brownian motion on fractals in chemical reactions (Northeast Regional Meeting, Binghamton, NY, October/5-8/16). Academia Journal of Scientific Research (ISSN 2315-7712) DOI: 10.15413/ajsr.2016.0168 In preparation for publication.

[6] Michael Fundator Applications of Multidimensional Time Model for Probability Cumulative Function for design and analysis of stepped wedge randomized trials. Academia Journal of Scientific Research (ISSN 2315-7712) DOI: 10.15413/ajsr.2016.0169 In preparation for publication.

[7] Michael Fundator Multidimensional Time Model for Probability Cumulative Function and Connections Between Deterministic Computations and Probabilities Journals of Mathematics and System Science 7 (2017) 101-109 doi: 10.17265/2159-5291/2017.04.001.

[8] N. Bohr, H. A. Kramers, J. C. Slater The quantum Theory of Radiation.

[9] J. Hendry Bohr-Kramers-Slater: A Virtual Theory of Virtual Oscillators and Its Role in the History of Quantum Mechanics. A. N.

[10] M. P. Bianchi A BGK-type model for a gas mixture undergoing reversible reaction.

[11] H. A. Kramers “Brownian motion in a field of force and the diffusion model of chemical reactions.” Physica, 7, 4, 284-304 (1940).

[12] R. von Mises “Probability, Statistics and Truth”.

[13] P. Morters “Lecture notes.”

[14] D. R. Brillinger “Moments, cumulants, and some applications to stationary random processes”.

[15] Y. L. Tong, “Relationship between stochastic inequalities and some classical mathematical inequalities,” Journal of Inequalities and Applications, vol. 1, no. 1, pp. 85-98, 1997.

[16] J. D Esary, F Proschan, D. W Walkup “Association of random variables with applications” Ann. Math. Statist., 38 (1976), pp. 1466–1474

[17] Paul Richard Halmos “Lectures on Boolean Algebras” Van Nostrand.

[18] Khinchin “Three pearls of Number theory”.

[19] V Blåsjö The Evolution of the Isoperimetric Problem https://www.maa.org/sites/default/files/pdf/upload_library/22/Ford/blasj526.pdf.

[20] Sariel Har-Peled The Johnson-Lindenstrauss Lemma.

[21] R. Tapia The Isoperimetric Problem Revisited: Extracting a Short Proof of Sufficiency from Euler's 1744 Approach to Necessity.

[22] P. J. H Pineyro Gergonne; The isoperimetric problem and the Steiner's symmetrization.

[23] Gergonne, J. D.: Goeometrie. Recherche de la surface plane de moindre contour, entre toutes celles de m’em,e etendue, et du corps de moindre surface, entre tous ceux de m’em,e volume, Annales de Gergonne 4 (1813-1814), 338-343.
[24] Hilbert D. “Über die Darstellung definer Formen als Summe von Formenquadraten” Math Ann. 32, 342-350 (188).

[25] I. Schur, Bemerkungen zur Theorie der beschränkten Bilinearformen mit endlich vielen Veränderlichen, J. reine angew. Math., 140 (11).

[26] D. V. Widder, An inequality related to one of Hilbert’s, J. London Math. Soc. 4, 194-198.

[27] Frazer C. Note on Hilber’s inequality, J. London Math. Soc. 21/46 pp. 7-9.

[28] G. J. O. Jameson Hilbert’s inequality and related results Notes

[29] G. H. Hardy, “Prolegomena to a chapter on inequalities.”

[30] G. H. Hardy, J. E. Littlewood and G. Pólya, Inequalities, 2nd ed., Cambridge University Press.

[31] Mitrinovic, Dragoslav S. Analytic Inequalities.

[32] G. D. Handler Hilbert and Hardy type inequalities.

[33] H. S. Wilf, Finite sections of some classical inequalities, Ergebnisse der Mathematik, Band 52.

[34] Minkowski, Hermann (1896). Geometrie der Zahlen. Leipzig: Teubner.

[35] Milman, Vitali D. (1986). "Inégalité de Brunn-Minkowski inverse et applications à la théorie locale des espaces normés. [An inverse form of the Brunn-Minkowski inequality, with applications to the local theory of normed spaces]." C. R. Acad. Sci. Paris Sér. I Math. 302 (1): 25–28.

[36] E. F. Beckenbach and R. Bellman. “Inequalities”, Springer, Berlin, 1983.

[37] Fan, K., O. Taussky and J. Todd. Discrete analogues of inequalities of Wirtinger. Monatsh. Math, 59 (1955), 73-90.

[38] C. Stein, “Inadmissibility of the Usual Estimator for the Mean of a Multivariate Normal Distribution,” in Proc. Third Berkeley Symp. On Math. Statist. and Prob., vol. 1, pp. 197-206.

[39] C. Stein. Estimation of the mean of a multivariate normal distribution. Ann. Stat., 9 (6): 1135–1151.

[40] D. E. A. Giles. et al The positive-part Stein-rule estimator and tests of hypotheses. Economics Letters 26 (1988): 49-52.

[41] Fortuin, C. M.; Kasteleyn, P. W.; Ginibre, J. Correlation inequalities on some partially ordered sets. Comm. Math. Phys. 22 (1971), no. 2, 89–103.

[42] R. A. L. Carter et al Unbiased Estimation of the MSE Matrix of Stein-Rule Estimators, Confidence Ellipsoids, and Hypothesis Testing.

[43] H. M. Hudson “A Natural Identity for exponential families with applications in multiparameter estimation” The Annals of Statistics V. 6, N 3.

[44] Z-G Xiao et al A simple new proof of Fan-Taussky-Todd inequalities.

[45] H. Alzer Converse of two inequalities of Ky Fan, O. Taussky, and J. Todd Journal of Mathematical Analysis and Applications Volume 161, Issue 1, October 91, 142-147 https://doi.org/10.1016/0022-247X(91)90365-7.