

Some asymptotic results for integrated empirical processes with applications to statistical tests

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London – December 16, 2017

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1. Settings

1. Settings — Sampling

Definitions –

- ① Let $F = \{F(t), t \in \mathbb{R}\}$ be a **continuous** distribution function [d.f.] and denote by $Q = \{Q(u), u \in [0, 1]\}$ the usual quantile function (generalized inverse) pertaining to F defined as

$$Q(u) := \inf\{t \in \mathbb{R} : F(t) \geq u\} \quad u \in (0, 1)$$

$$Q(0) := \lim_{u \downarrow 0} Q(u) \quad \text{and} \quad Q(1) := \lim_{u \uparrow 1} Q(u)$$

- ② Let $\{U_i, i \in \mathbb{N}^*\}$ be a sequence of i.i.d. random variables [r.v.] uniformly distributed on $[0, 1]$.

- Set for any $i \in \mathbb{N}^*$:

$$X_i := Q(U_i)$$

The **sample** $\{X_i, i \in \mathbb{N}^*\}$ consists of i.i.d. r.v.'s with d.f. F :
 $F(t) = \mathbb{P}\{X_1 \leq t\}$ for $t \in \mathbb{R}$.

- Conversely:

$$U_i = F(X_i)$$

1. Settings — Empirical d.f./processes

Definitions –

- ③ **Empirical d.f.'s** based upon the samples X_1, \dots, X_n and U_1, \dots, U_n : for each $n \in \mathbb{N}^*$,

$$\mathbb{F}_n(t) := \frac{1}{n} \#\{i \in \{1, \dots, n\} : X_i \leq t\} = \frac{1}{n} \sum_{i=1}^n \mathbb{1}_{\{X_i \leq t\}} \quad t \in \mathbb{R}$$

$$\mathbb{U}_n(u) := \frac{1}{n} \#\{i \in \{1, \dots, n\} : U_i \leq u\} = \frac{1}{n} \sum_{i=1}^n \mathbb{1}_{\{U_i \leq u\}} \quad u \in [0, 1]$$

- ④ Related **empirical processes**:

$$\alpha_n(t) := \sqrt{n} (\mathbb{F}_n(t) - F(t)) \quad t \in \mathbb{R}$$

$$\beta_n(u) := \sqrt{n} (\mathbb{U}_n(u) - u) \quad u \in [0, 1]$$

Proposition –

$$\alpha_n(t) = \beta_n(F(t)) \quad t \in \mathbb{R}$$

→ It is enough to work with the uniform law (i.e. the U_i 's, \mathbb{U}_n, β_n)

1. Settings — Empirical d.f./processes

Definitions –

- 5 Families of *integrated d.f.'s* and *integrated empirical d.f.'s* associated with the uniform d.f., for any $n \in \mathbb{N}^*$, $u \in [0, 1]$:

$$\begin{aligned} U^{(0)}(u) &:= u & U_n^{(0)}(u) &:= U_n(u) \\ U^{(1)}(u) &:= \int_0^u v \, dv & U_n^{(1)}(u) &:= \int_0^u U_n(v) \, dU_n(v) \end{aligned}$$

→ *Introduced by Henze & Nikitin – 2003*

1. Settings — Empirical d.f./processes

Definitions –

- 5 Families of *integrated d.f.'s* and *integrated empirical d.f.'s* associated with the uniform d.f., for any $n \in \mathbb{N}^*$, $u \in [0, 1]$:

$$\begin{aligned} U^{(0)}(u) &:= u & U_n^{(0)}(u) &:= U_n(u) \\ U^{(1)}(u) &:= \int_0^u v \, dv & U_n^{(1)}(u) &:= \int_0^u U_n(v) \, dU_n(v) \end{aligned}$$

and for any $p \in \mathbb{N}^*$:

$$\begin{aligned} U^{(p)}(u) &:= \int_0^u dv_1 \int_0^{v_1} dv_2 \dots \int_0^{v_{p-1}} v_p \, dv_p \\ U_n^{(p)}(u) &:= \int_0^u dU_n(v_1) \int_0^{v_1} dU_n(v_2) \dots \int_0^{v_{p-1}} U_n(v_p) \, dU_n(v_p) \end{aligned}$$

- 6 Family of *integrated empirical processes*:

$$\beta_n^{(p)}(u) := \sqrt{n} \left(U_n^{(p)}(u) - U^{(p)}(u) \right)$$

1. Settings — Empirical d.f./processes

Proposition – For each $p \in \mathbb{N}$:

- **Integrated d.f.'s:**

$$U^{(p)}(u) = \frac{u^{p+1}}{(p+1)!}$$

- **Integrated empirical d.f.'s:**

$$\begin{aligned} U_n^{(p)}(u) &= \frac{1}{n^{p+1}} \binom{nU_n(u) + p}{p+1} \quad \text{a.s.} \\ &= \frac{U_n(u)^{p+1}}{(p+1)!} + \sum_{k=1}^p \left[\begin{matrix} p+1 \\ k+1 \end{matrix} \right] \frac{U_n(u)^k}{n^{p-k+1}} \quad n \in \mathbb{N}^* \quad u \in [0, 1] \end{aligned}$$

where $\left[\begin{matrix} p+1 \\ k+1 \end{matrix} \right]$, $1 \leq k \leq p$, are the unsigned Stirling numbers:

$$\prod_{i=1}^p (x+i) = \sum_{k=0}^p \left[\begin{matrix} p+1 \\ k+1 \end{matrix} \right] x^k$$

1. Settings — Some Gaussian processes

Definitions –

- ① **Wiener process** $\mathbb{W} = \{\mathbb{W}(s), s \geq 0\}$:

$$\mathbb{E}(\mathbb{W}(s)\mathbb{W}(t)) = s \wedge t \quad s, t \geq 0$$

- ② **Brownian bridge** $\mathbb{B} = \{\mathbb{B}(u), u \in [0, 1]\}$:

$$\mathbb{E}(\mathbb{B}(u)\mathbb{B}(v)) = u \wedge v - uv \quad u, v \in [0, 1]$$

- ③ **Kiefer process** $\mathbb{K} = \{\mathbb{K}(s, u), s \geq 0, u \in [0, 1]\}$:

$$\mathbb{E}(\mathbb{K}(s, u)\mathbb{K}(t, v)) = (s \wedge t)(u \wedge v - uv) \quad s, t \geq 0 \quad u, v \in [0, 1]$$

Recall – ‘Extracted’ Kiefer process $\{\mathbb{K}(n, u), n \in \mathbb{N}^*, u \in [0, 1]\}$:

$$\mathbb{K}(n, u) = \sum_{i=1}^n \mathbb{B}_i(u) \quad n \in \mathbb{N}^* \quad u \in [0, 1]$$

where $\{\mathbb{B}_i, i \in \mathbb{N}^*\}$ is a sequence of independent Brownian bridges

1. Settings — Aims/Motivations

Aims –

- ① To derive ***strong approximations*** for ***integrated empirical processes*** by a family of Gaussian processes involving Brownian bridge and Kiefer process similar to the famous ***'Hungarian construction'***
 - ② To provide ***new statistical tests:***
 - Kolmogorov-Smirnov-like tests
 - Cramér-von Mises-like tests
 - Change-point detection, etc.
-

Motivations –

- ① Study of ***integrated Brownian motions*** and ***bridges***
- ② Computation of ***Bahadur-efficiency*** for various tests
 - Leading to ***more performant procedures...***

2. Asymptotics

2. Asymptotics — ‘Hungarian construction’

Theorem – **Gaussian approximations** (Komlós–Major–Tusnády 1975)

- ① On a suitable probability space, we may define the uniform empirical process $\{\beta_n, n \in \mathbb{N}^*\}$, in combination with a sequence of Brownian bridges $\{\mathbb{B}_n, n \in \mathbb{N}^*\}$, such that

$$\sup_{u \in [0,1]} |\beta_n(u) - \mathbb{B}_n(u)| \underset{n \rightarrow \infty}{=} O\left(\frac{\log n}{\sqrt{n}}\right) \text{ a.s.}$$

- ② On a suitable probability space, we may define the uniform empirical process $\{\beta_n, n \in \mathbb{N}^*\}$, in combination with a Kiefer process $\{\mathbb{K}(s, u), s \geq 0, u \in [0, 1]\}$, such that

$$\max_{1 \leq k \leq n} \sup_{u \in [0,1]} |\sqrt{k} \beta_k(u) - \mathbb{K}(k, u)| \underset{n \rightarrow \infty}{=} O((\log n)^2) \text{ a.s.}$$

2. Asymptotics — Gaussian approximations

Theorem 1 – Gaussian approximation 1 (AA-B-L 2017)

On a suitable probability space, we may define the p -fold integrated uniform empirical process $\{\beta_n^{(p)}, n \in \mathbb{N}^*\}$, in combination with a sequence of Brownian bridges $\{\mathbb{B}_n, n \in \mathbb{N}^*\}$, such that

$$\sup_{u \in [0,1]} |\beta_n^{(p)}(u) - \mathbb{B}_n^{(p)}(u)| \underset{n \rightarrow \infty}{=} O\left(\frac{\log n}{\sqrt{n}}\right) \quad \text{a.s.}$$

where $\mathbb{B}_n^{(p)}$ is the process defined by

$$\mathbb{B}_n^{(p)}(u) := \frac{1}{p!} u^p \mathbb{B}_n(u) \quad u \in [0, 1]$$

2. Asymptotics — Gaussian approximations

Theorem 2 – Gaussian approximation 2 (AA-B-L 2017)

On a suitable probability space, we may define the p -fold integrated uniform empirical process $\{\beta_n^{(p)}, n \in \mathbb{N}^*\}$, in combination with a Kiefer process $\{\mathbb{K}(s, u), s \geq 0, u \in [0, 1]\}$, such that

$$\max_{1 \leq k \leq n} \sup_{u \in [0, 1]} \left| \sqrt{k} \beta_k^{(p)}(u) - \mathbb{K}^{(p)}(k, u) \right| \underset{n \rightarrow \infty}{=} O((\log n)^2) \quad \text{a.s.}$$

where $\mathbb{K}^{(p)}$ is the process defined by

$$\mathbb{K}^{(p)}(s, u) := \frac{1}{p!} u^p \mathbb{K}(s, u) \quad s \geq 0 \quad u \in [0, 1]$$

2. Asymptotics — Gaussian approximations

Sketch of proof –

Lemma 1 – Representation of $\beta_n^{(p)}$ by means of β_n and \mathbb{U}_n

$$\beta_n^{(p)}(u) = \frac{1}{p!} u^p \beta_n(u) + \sum_{k=2}^{p+1} \frac{\binom{p+1}{k}}{(p+1)!} \frac{u^{p+1-k}}{n^{(k-1)/2}} \beta_n(u)^k \\ + \sum_{k=1}^p \left[\begin{matrix} p+1 \\ k+1 \end{matrix} \right] \frac{\mathbb{U}_n(u)^k}{n^{p-k+1/2}} \quad n \in \mathbb{N}^* \quad u \in [0, 1]$$

Lemma 2 – Chung's law of the iterated logarithm for β_n

$$\sup_{u \in [0,1]} |\beta_n(u)| \underset{n \rightarrow \infty}{=} O(\sqrt{\log \log n}) \quad \text{a.s.}$$

2. Asymptotics — Gaussian approximations

Corollary – Law of the iterated logarithm for $\beta_n^{(p)}$

$$\limsup_{n \rightarrow \infty} \frac{\sup_{u \in [0,1]} |\beta_n^{(p)}(u)|}{\sqrt{\log \log n}} = \frac{(p+1/2)^{p+1/2}}{p! (p+1)^{p+1}} \quad \text{a.s.}$$

Examples –

$$\limsup_{n \rightarrow \infty} \frac{\sup_{u \in [0,1]} |\beta_n^{(0)}(u)|}{\sqrt{\log \log n}} = \frac{1}{\sqrt{2}} \quad \text{a.s.} \quad \text{Chung's LIL}$$

$$\limsup_{n \rightarrow \infty} \frac{\sup_{u \in [0,1]} |\beta_n^{(1)}(u)|}{\sqrt{\log \log n}} = \frac{3}{8} \sqrt{\frac{3}{2}} \quad \text{a.s.}$$

2. Asymptotics — A key inequality

Complement – An exponential bound (AA-B-L 2017)

For any $n \in \mathbb{N}^*$ and large enough x :

$$\mathbb{P} \left\{ \sup_{u \in [0,1]} |\beta_n^{(p)}(u) - \mathbb{B}_n^{(p)}(u)| \geq \frac{1}{\sqrt{n}} (x + 12 \log(n)) \right\} \leq A_p \sum_{k=2}^{p+1} e^{-B_p x^{2/k} n^{1-2/k}}$$

where A_p and B_p are positive constants depending on p

→ Supplies Theorem 1 and other consequences...

3. Applications

3. Applications — Goodness-of-fit

Problem 1 – Testing the null hypothesis (fitting a distribution)

$$\mathcal{H}_0: F = F_0$$

Tools: some statistics

Definitions –

① *Kolmogorov-Smirnov-like statistic*

$$S_n^{(p)} := \sqrt{n} \sup_{t \in \mathbb{R}} \left| \left(\mathbb{F}_n^{(p)}(t) - F_0^{(p)}(t) \right) \right|$$

② *Cramér-von Mises-like statistic*

$$T_n^{(p)} := n \int_{\mathbb{R}} \left(\mathbb{F}_n^{(p)}(t) - F_0^{(p)}(t) \right)^2 dF_0(t)$$

3. Applications — Goodness-of-fit

Problem 1 – Testing the null hypothesis (*fitting a distribution*)

$$\mathcal{H}_0: F = F_0$$

Theorem 3 – (AA-B-L 2017)

Under \mathcal{H}_0 :

$$\left| \mathbf{S}_n^{(p)} - \sup_{u \in [0,1]} |\mathbb{B}_n^{(p)}(u)| \right| \stackrel{p}{\underset{q}{\underset{n \rightarrow \infty}{\Rightarrow}}} O\left(\frac{\log n}{\sqrt{n}}\right) \text{ a.s.}$$

$$\left| \mathbf{T}_n^{(p)} - \int_0^1 [\mathbb{B}_n^{(p)}(u)]^2 du \right| \stackrel{p}{\underset{q}{\underset{n \rightarrow \infty}{\Rightarrow}}} O\left(\sqrt{\frac{\log \log n}{n}} \log n\right) \text{ a.s.}$$

3. Applications — Goodness-of-fit

Problem 1 – Hypothesis testing (*fitting a distribution*)

$$\mathcal{H}_0: F = F_0 \quad \text{versus} \quad \mathcal{H}_1: F = F_1$$

Simulations – Tests for uniformity: $F_0(u) = u$

- Random samples of size $n = 100$ at level $\alpha = 0.05$ based on 10000 replications for each $p \in \{0, 1, 2, 3\}$
- Computation of powers

Various alternatives

$$A_k: F_1(u) = 1 - (1 - u)^k \quad u \in [0, 1]$$

$$B_k: F_1(u) = \begin{cases} 2^{k-1} u^k & u \in [0, 0.5] \\ 1 - 2^{k-1} (1 - u)^k & u \in [0.5, 1] \end{cases}$$

$$C_k: F_1(u) = \begin{cases} 0.5 - 2^{k-1} (0.5 - u)^k & u \in [0, 0.5] \\ 0.5 + 2^{k-1} (u - 0.5)^k & u \in [0.5, 1] \end{cases}$$

F_1	$S_n^{(0)}$	$S_n^{(1)}$	$S_n^{(2)}$	$S_n^{(3)}$
$A_{1.5}$	0.91	0.96	0.94	0.58
A_2	1	1	1	0.99
$B_{1.5}$	0.34	0.54	0.61	0.17
B_2	0.94	0.97	0.98	0.76
B_3	1	1	1	0.99
$C_{1.5}$	0.48	0.34	0.31	0.48
C_2	0.96	0.81	0.75	0.86
C_3	1	0.99	0.99	0.99

3. Applications — Two-sample problem

Problem 2 – Testing the null hypothesis (*comparison of two samples*)

$$\mathcal{H}'_0: F = G$$

Settings –

- X_1, \dots, X_m and Y_1, \dots, Y_n : independent random **samples** from **continuous** d.f.'s F and G
- $\mathbb{F}_m^{(p)}$ and $\mathbb{G}_n^{(p)}$: their **p -fold integrated empirical d.f.'s**

Definitions –

- 1 **Kolmogorov-Smirnov-like statistic**

$$S_{m,n}^{(p)} := \sqrt{\frac{mn}{m+n}} \sup_{t \in \mathbb{R}} \left| \left(\mathbb{F}_m^{(p)}(t) - \mathbb{G}_n^{(p)}(t) \right) \right|$$

- 2 **Cramér-von Mises-like statistic**

$$T_{m,n}^{(p)} := \frac{mn}{m+n} \int_{\mathbb{R}} \left(\mathbb{F}_m^{(p)}(t) - \mathbb{G}_n^{(p)}(t) \right)^2 dF(t)$$

3. Applications — Two-sample problem

Problem 2 – Testing the null hypothesis (*comparison of two samples*)

$$\mathcal{H}'_0: F = G$$

Theorem 4 – (AA-B-L 2017)

Under \mathcal{H}'_0 :

$$\left| \mathbf{S}_{m,n}^{(\rho)} - \sup_{u \in [0,1]} \left| \mathbb{B}_{m,n}^{(\rho)}(u) \right| \right|_{m,n \rightarrow \infty} = O(\varphi(m,n)) \quad \text{a.s.}$$

$$\left| \mathbf{T}_{m,n}^{(\rho)} - \int_0^1 \left[\mathbb{B}_{m,n}^{(\rho)}(u) \right]^2 du \right|_{m,n \rightarrow \infty} = O(\phi(m,n)) \quad \text{a.s.}$$

where

$$\varphi(m,n) := \max \left(\frac{\log m}{\sqrt{m}}, \frac{\log n}{\sqrt{n}} \right)$$

$$\phi(m,n) := \max \left(\sqrt{\frac{\log \log m}{m}} \log m, \sqrt{\frac{\log \log n}{n}} \log n \right)$$

3. Applications — Two-sample problem

Problem 2 – Testing the null hypothesis (*comparison of two samples*)

$$\mathcal{H}'_0: F = G$$

Theorem 4 – (AA-B-L 2017)

Under \mathcal{H}'_0 :

$$\left| \mathbf{S}_{m,n}^{(\rho)} - \sup_{u \in [0,1]} |\mathbb{B}_{m,n}^{(\rho)}(u)| \right|_{m,n \rightarrow \infty} = O(\varphi(m,n)) \quad \text{a.s.}$$

$$\left| \mathbf{T}_{m,n}^{(\rho)} - \int_0^1 [\mathbb{B}_{m,n}^{(\rho)}(u)]^2 du \right|_{m,n \rightarrow \infty} = O(\phi(m,n)) \quad \text{a.s.}$$

where

$$\mathbb{B}_{m,n}^{(\rho)}(u) := \frac{1}{\rho!} u^\rho \left(\sqrt{\frac{n}{m+n}} \mathbb{B}_m^1(u) - \sqrt{\frac{m}{m+n}} \mathbb{B}_n^2(u) \right) \quad u \in [0, 1]$$

$\{\mathbb{B}_n^1, n \in \mathbb{N}^*\}$ and $\{\mathbb{B}_n^2, n \in \mathbb{N}^*\}$ being two independent sequences of Brownian bridges

3. Applications — Change-point problem

Problem 3 – Testing the null hypothesis (*homogeneity of the sample*)

$$\mathcal{H}_0'' : X_1, \dots, X_n \text{ have d.f. } F_0$$

Against the alternative hypothesis:

$$\mathcal{H}_1'' : \exists k^* \in \{1, \dots, n-1\}, \begin{cases} X_1, \dots, X_{k^*} \text{ have d.f. } F_0 \\ X_{k^*+1}, \dots, X_n \text{ have d.f. } F_1 \end{cases}$$

Settings –

- $\mathbb{F}_k^{(p)-}$: p -fold integrated empirical d.f.
based upon the k first observations
- $\mathbb{F}_{n-k}^{(p)+}$: p -fold integrated empirical d.f.
based upon the $(n - k)$ last observations

3. Applications — Change-point problem

Problem 3 – Testing the null hypothesis (*homogeneity of the sample*)

$$H_0'' : X_1, \dots, X_n \text{ have d.f. } F_0$$

Definitions –

- **Ad hoc empirical process**

$$\tilde{\alpha}_n^{(p)}(s, t) := \frac{\lfloor ns \rfloor (n - \lfloor ns \rfloor)}{n^{3/2}} \left(\mathbb{F}_{\lfloor ns \rfloor}^{(p)-}(t) - \mathbb{F}_{n - \lfloor ns \rfloor}^{(p)+}(t) \right) \quad s \in (0, 1) \quad t \in \mathbb{R}$$

- **'Hybrid' Kiefer process**

For $\{\mathbb{K}_1(s, u), s \in \mathbb{R}, u \in [0, 1]\}$ and $\{\mathbb{K}_2(s, u), s \in \mathbb{R}, u \in [0, 1]\}$ two independent Kiefer processes, set

$$\mathring{\mathbb{K}}_n(s, u) := \begin{cases} \frac{1}{\sqrt{n}} \left[\mathbb{K}_2(\lfloor ns \rfloor, u) - s(\mathbb{K}_1(\lfloor n/2 \rfloor, u) + \mathbb{K}_2(\lfloor n/2 \rfloor, u)) \right] & s \in [0, \frac{1}{2}] \quad u \in [0, 1] \\ \frac{1}{\sqrt{n}} \left[-\mathbb{K}_1(\lfloor n(1-s) \rfloor, u) + (1-s)(\mathbb{K}_1(\lfloor n/2 \rfloor, u) + \mathbb{K}_2(\lfloor n/2 \rfloor, u)) \right] & s \in [\frac{1}{2}, 1] \quad u \in [0, 1] \end{cases}$$

3. Applications — Change-point problem

Problem 3 – Testing the null hypothesis (*homogeneity of the sample*)

$$\mathcal{H}_0'' : X_1, \dots, X_n \text{ have d.f. } F_0$$

Theorem 5 – *Strong approximation* (AA–B–L 2017)

On a suitable probability space, it is possible to define

$\{\tilde{\alpha}_n^{(p)}, n \in \mathbb{N}^*\}$, jointly with two independent Kiefer processes $\{\mathbb{K}_1(\mathbf{s}, \mathbf{u}), \mathbf{s} \in \mathbb{R}, \mathbf{u} \in [0, 1]\}$ and $\{\mathbb{K}_2(\mathbf{s}, \mathbf{u}), \mathbf{s} \in \mathbb{R}, \mathbf{u} \in [0, 1]\}$, such that, under \mathcal{H}_0'' :

$$\sup_{\substack{\mathbf{s} \in (0, 1) \\ t \in \mathbb{R}}} |\tilde{\alpha}_n^{(p)}(\mathbf{s}, t) - \overset{\circ}{\mathbb{K}}_n^{(p)}(\mathbf{s}, F_0(t))| \underset{n \rightarrow \infty}{=} O\left(\frac{(\log n)^2}{\sqrt{n}}\right) \text{ a.s.}$$

where

$$\overset{\circ}{\mathbb{K}}_n^{(p)}(\mathbf{s}, \mathbf{u}) := \frac{1}{p!} u^p \overset{\circ}{\mathbb{K}}_n(\mathbf{s}, \mathbf{u}) \quad n \in \mathbb{N}^* \quad \mathbf{s}, \mathbf{u} \in [0, 1]$$

4. Further investigations

4. Further investigations

1 Weighted approximations

→ **problem:**
$$\sup_{u \in (0,1)} \frac{|\beta_n^{(p)}(u) - \mathbb{B}_n^{(p)}(u)|}{u^\gamma} \underset{n \rightarrow \infty}{=} O(?)$$

2 Poisson approximation

Let $\{\nu_n, n \in \mathbb{N}^*\}$ be a sequence of Poisson r.v.'s independent of the sequence $\{X_i, i \in \mathbb{N}^*\}$ such that $\mathbb{E}(\nu_n) = n$. Introduce the Poisson bridges $\{\mathbb{P}_n, n \in \mathbb{N}^*\}$ defined, for each $n \in \mathbb{N}^*$, by

$$\mathbb{P}_n(u) := \frac{1}{\sqrt{n}} \left(\sum_{i=1}^{\nu_n} \mathbb{1}_{\{U_i \leq u\}} - \nu_n u \right) \quad u \in [0, 1]$$

→ **problem:**
$$\sup_{u \in [0,1]} |\beta_n^{(p)}(u) - \mathbb{P}_n^{(p)}(u)| \underset{n \rightarrow \infty}{=} O(?)$$

3 Higher dimension, multivariate case...

Thank you for your attention!

References:

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Slides available at http://math.univ-lyon1.fr/~alachal/exposes/slides_london_2017.pdf