

EMPIRICAL PROCESSES, VAPNIK-CHERVONENKIS CLASSES AND POISSON PROCESSES

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Abstract. For background of this paper* see [2]. Given a probability space (X, \mathcal{A}, P) , let G_P be the Gaussian process with mean 0, indexed by \mathcal{A} , and such that

$$EG_P(A)G_P(B) = P(A \cap B) - P(A)P(B), \quad A, B \in \mathcal{A}.$$

(1) Let $\mathcal{C} \subset \mathcal{A}$ and suppose that, for all probability measures (laws) Q on \mathcal{A} , G_Q has a version with bounded sample functions on \mathcal{C} . (For example, suppose \mathcal{C} is a "universal Donsker class".) Then, for some n , no set F of n elements has all its subsets of the form $C \cap F$, $C \in \mathcal{C}$, i.e. \mathcal{C} is a Vapnik-Chervonenkis class. An example shows that limit theorems for empirical measures need not hold uniformly over a Vapnik-Chervonenkis class of measurable sets, unless further measurability is assumed.

(2) For a law P on $X = \{1, 2, \dots\}$, the collection 2^X of all subsets is a Donsker class if and only if

$$\sum_m P(m)^{1/2} < \infty.$$

(3) For any probability space (X, \mathcal{A}, P) , suppose \mathcal{C} is a P -Donsker class, $\mathcal{C} \subset \mathcal{A}$. Let T_a be a Poisson point process with intensity measure aP , $a > 0$. Then, as $a \rightarrow \infty$, $(T_a - aP)/a^{1/2}$ converges in law, with respect to uniform convergence on \mathcal{C} , to the Gaussian process W_P with mean 0 and $EW_P(A)W_P(B) = P(A \cap B)$, $A, B \in \mathcal{C}$.

1. Introduction. Let (X, \mathcal{A}, P) be any probability space. Let G_P and W_P be the Gaussian processes, indexed by \mathcal{A} , with mean 0 and such that for all $A, B \in \mathcal{A}$

$$EW_P(A)W_P(B) = P(A \cap B) \quad \text{and} \quad EG_P(A)G_P(B) = P(A \cap B) - P(A)P(B).$$

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Then for all $A \in \mathcal{A}$ we can write

$$W_P(A) = G_P(A) + P(A)H,$$

where $H := W_P(X)$ is a standard Gaussian variable independent of G_P .

Let X_1, X_2, \dots be independent and identically distributed with law P , and let P_n be the random empirical measure $n^{-1}(\delta_{X_1} + \dots + \delta_{X_n})$. Let $\mathcal{C} \subset \mathcal{A}$. In [2], \mathcal{C} was called a *P-Donsker class* if the convergence of laws $\mathcal{L}(n^{1/2}(P_n - P)) \rightarrow \mathcal{L}(G_P)$ holds with respect to uniform convergence on \mathcal{C} in a suitable sense, together with some measurability conditions. Here we will need only the following Skorohod-Wichura form of convergence (see [2], p. 900-902):

1.1. If \mathcal{C} is a *P-Donsker class*, then there is a probability space $(\Omega, \mathcal{B}, \Pr)$ and for $n = 1, 2, \dots$ there are processes $(\omega, C) \rightarrow A_n(\omega, C)$, $\omega \in \Omega$, $C \in \mathcal{C}$, such that, for each fixed n , the laws of the processes $n^{1/2}(P_n - P)$ and A_n are the same and such that

$$\limsup_{n \rightarrow \infty} \sup_{C \in \mathcal{C}} |A_n(\omega, C) - G_P(C)(\omega)| = 0 \text{ a.s.,}$$

where G_P is defined on the probability space Ω . It follows that

$$\sup_{C \in \mathcal{C}} |G_P(C)(\omega)| < \infty \text{ a.s.}$$

Sections 2, 3 and 4 use the above, but are independent of one another.

2. Universal Donsker classes are Vapnik-Chervonenkis classes. For any set X let 2^X be the collection of all its subsets (power set). Let $\mathcal{C} \subset 2^X$. Then \mathcal{C} is said to *shatter* a set $F \subset X$ if $2^F = \{F \cap C : C \in \mathcal{C}\}$. Also, \mathcal{C} is called a *Vapnik-Chervonenkis class* if, for some finite n , no set F with n elements is shattered by \mathcal{C} .

2.1. THEOREM. For any set X and collection \mathcal{C} of subsets of X which is not a Vapnik-Chervonenkis class, there are a purely atomic probability measure P on X and a countable collection $\mathcal{D} \subset \mathcal{C}$ such that G_P is almost surely unbounded on \mathcal{D} .

Proof. Since \mathcal{C} shatters sets of all sizes, for each $n = 1, 2, \dots$ there is a set F_n with 4^n elements, shattered by \mathcal{C} . Let

$$G_n := F_n \setminus \bigcup_{j < n} F_j.$$

Then the G_n are disjoint and have cardinality

$$\text{card}(G_n) \geq 4^n - \sum_{j=1}^{n-1} 4^j = 4^n - (4^n - 4)/3 > 2^n,$$

with G_n shattered by \mathcal{C} . Take $E_n \subset G_n$ with $\text{card}(E_n) = 2^n$. Then E_n remain disjoint and are shattered by \mathcal{C} .

Let $P(\{x\}) = 6/(\pi^2 n^2 \cdot 2^n)$ for each $x \in E_n$, and let $P = 0$ outside $\bigcup_{n=1}^{\infty} E_n$. Then P is a purely atomic probability measure on X .

Let \mathcal{D} be a countable subset of \mathcal{C} which shatters each of the E_n . Let us fix n . Then, for each $C \in \mathcal{D}$,

$$W_P(C) = W_P(C \cap E_n) + W_P(C \setminus E_n).$$

Thus for any $K, 0 < K < \infty$, we have

$$\{\omega: |W_P(C)(\omega)| < K \text{ for all } C \in \mathcal{D}\} \subset \mathcal{E}_1 \cup \mathcal{E}_2,$$

where

$$\mathcal{E}_1 := \{\omega: |W_P(B)(\omega)| \leq 2K \text{ for all } B \subset E_n\},$$

$$\mathcal{E}_2 := \{\omega: \text{for some } B \subset E_n, |W_P(B)(\omega)| > 2K, \text{ and for all such } B \text{ and all } C \in \mathcal{D} \text{ with } C \cap E_n = B \text{ we have } |W_P(C \setminus E_n)(\omega)| > K\}.$$

Let

$$S_n := \sum_{x \in E_n} |W_P(\{x\})|.$$

Then since $\sup\{|W_P(B)|: B \subset E_n\} \geq S_n/2$, we have $\mathcal{E}_1 \subseteq \{S_n \leq 4K\}$. For each $x \in E_n$, $W_P(\{x\})$ is a normal random variable with mean 0 and variance $\sigma_n^2 := 6/(\pi^2 n^2 \cdot 2^n)$. Thus

$$E|W_P(\{x\})| = (2/\pi)^{1/2} \sigma_n \quad \text{and} \quad \text{var}(|W_P(\{x\})|) = \sigma_n^2(1 - 2/\pi).$$

Then

$$E S_n = 2^n (2/\pi)^{1/2} \sigma_n \quad \text{and} \quad \text{var}(S_n) = (6/(\pi^2 n^2))(1 - 2/\pi),$$

since W_P has independent values on disjoint sets. Hence, by Chebyshev's inequality, for large n we get

$$\begin{aligned} \Pr\{S_n \leq 4K\} &\leq \Pr\{|S_n - E S_n| \geq E S_n - 4K\} \leq n^{-2}/(4K - E S_n)^2 \\ &\leq 1/(4Kn - 2^{n/2} (12/\pi^3)^{1/2})^2 := f(n, K) \rightarrow 0 \quad \text{as } n \rightarrow \infty \end{aligned}$$

for any fixed K .

Now we consider the event \mathcal{E}_2 . Let $tt(n) := 2^{2^n}$. Enumerate 2^{E_n} by $B(1), \dots, B(tt(n))$, and let

$$M_1 := \{\omega: |W_P(B(1))(\omega)| > 2K\},$$

$$M_m := \{\omega \notin \bigcup_{j=1}^{m-1} M_j: |W_P(B(m))(\omega)| > 2K\}, \quad m \geq 2,$$

$$D_j := \{\omega \in M_j: \text{for all } C \in \mathcal{D} \text{ such that } C \cap E_n = B(j), |W_P(C \setminus E_n)(\omega)| > K\}.$$

By the independence of W_P on disjoint sets, we have

$$\Pr(D_j) = \Pr(M_j) \Pr\{\text{for all } C \in \mathcal{D} \text{ such that } C \cap E_n = B(j),$$

$$|W_P(C \setminus E_n)(\omega)| > K\} \leq \Pr(M_j) \cdot 2\Phi(-K),$$

where Φ is the standard normal distribution function, since, for any fixed set A , $W_P(A)$ is normal with mean 0 and variance less than 1. Now,

$$\mathcal{E}_2 \subset \bigcup_{1 \leq j \leq t(n)} D_j,$$

so that

$$\begin{aligned} \Pr(\mathcal{E}_2) &\leq \sum_{1 \leq j \leq t(n)} \Pr(M_j) \cdot 2\Phi(-K) \\ &= 2\Phi(-K) \Pr(|W_P(B)| > 2K \text{ for some } B \subset E_n) \leq 2\Phi(-K). \end{aligned}$$

It follows that

$$\Pr(|W_P(C)| < K \text{ for all } C \in \mathcal{D}) \leq f(n, K) + 2\Phi(-K).$$

Making K large enough, and then n large enough, completes the proof.

It follows that if \mathcal{C} is a *universal Donsker class*, i.e. it is a P -Donsker class for all P on the σ -algebra $\mathcal{A} \supset \mathcal{C}$, then \mathcal{C} is a Vapnik-Chervonenkis class. In [2], Section 7 and Correction, it is shown that every Vapnik-Chervonenkis class satisfying some measurability conditions is a universal Donsker class. The remaining problem is to find what measurability conditions are needed. The following example shows that some further measurability is necessary.

2.2. PROPOSITION. *There exist a set X and a class \mathcal{C} of countable subsets of X , which shatters no 2-element set, and a probability measure P such that almost surely*

$$\sup_{A \in \mathcal{C}} (P_n - P)(A) = 1 \quad \text{for all } n.$$

Assuming the continuum hypothesis, we can take $X = [0, 1]$ and P to be Lebesgue measure.

Proof. Let $(X, <)$ be an uncountable well-ordered set such that all its initial segments $\{x: x < y\}$, $y \in X$, are countable. Let \mathcal{C} be the collection of all these initial segments. Then \mathcal{C} does not shatter any set with two elements. Let P be any probability measure on X which is 0 on countable sets and 1 on their complements. Given any finite set $\{X_1, \dots, X_n\} \subset X$, there is a set A in \mathcal{C} containing all the X_i , so $(P_n - P)(A) = 1$, which completes the proof.

Steele [3] assumes that all sets in \mathcal{C} are measurable and that $\sup_{A \in \mathcal{C}} |(P_n - P)(A)|$ is measurable. These conditions are both satisfied in the example above. Thus it appears that further measurability conditions need to be added to some of the statements and proofs in [3].

3. When is 2^X P -Donsker for X countable? Let X be a countable set, say $X = \{1, 2, \dots\}$, and let P be a law on X with $P\{m\} := p_m$, $m = 1, 2, \dots$

3.1. THEOREM. *The collection 2^X of all subsets of X is a P -Donsker class if and only if*

$$(*) \quad \sum_m p_m^{1/2} < \infty.$$

Proof. Suppose $(*)$ holds. We have $E(v_n\{m\})^2 = p_m - p_m^2$ for all n and m , where $v_n := n^{1/2}(P_n - P)$. Thus $E|v_n\{m\}| \leq p_m^{1/2}$, and

$$\sup_n E \sum_{j \geq m} |v_n\{j\}| \rightarrow 0 \quad \text{as } m \rightarrow \infty.$$

So, for any $\varepsilon > 0$,

$$\sup_n \Pr \left\{ \sum_{j \geq m} |v_n\{j\}| > \varepsilon \right\} \rightarrow 0 \quad \text{as } m \rightarrow \infty.$$

Thus condition (b) in Theorem 1.2 of [2] holds; as the other conditions also hold for $\mathcal{C} = 2^X$, X countable, it is a P -Donsker class.

On the other hand, if $\sum_m p_m^{1/2} = \infty$, then

$$\sum_m |G_P\{m\}| = \infty \text{ a.s.}$$

by Proposition 6.6 of [1], letting $b_m := p_m^{1/2}$, $\varphi_m = 1_{\{m\}}/p_m^{1/2}$, and recalling the relations $L(1_A) = W_P(A) = G_P(A) + P(A)W_P(X)$. Thus G_P has sample functions almost surely unbounded on 2^X (it is enough to consider the countable collection of finite sets). Consequently, 2^X is not a Donsker class, which completes the proof.

4. A limit theorem for Poisson processes. Let (X, \mathcal{A}, μ) be a σ -finite measure space. Then the *Poisson process* T_μ with *intensity measure* μ is indexed by the measurable sets A with $\mu(A) < \infty$; $T_\mu(A)$ is a Poisson variable with parameter $\mu(A)$, and T_μ has independent values on disjoint sets, being additive for (finitely many) disjoint sets. These conditions, as is known, consistently define a stochastic process.

For $0 < \mu(X) < \infty$, let $T(X) = n$ be a Poisson variable with parameter $\mu(X)$. Then let

$$T = \sum_{1 \leq i \leq n} \delta_{X_i},$$

where the X_i are independent and identically distributed with law $\mu/\mu(X)$, and independent of n . It is easily seen that this T is a Poisson process T_μ .

Now let P be a probability measure and $0 < \lambda < \infty$. Then, as $\lambda \rightarrow \infty$, $(T_{\lambda P} - \lambda P)/\lambda^{1/2}$ converges in law to W_P , at least on any finite collection of measurable sets. For $\mathcal{C} \subset \mathcal{A}$, we say that this *convergence in law holds with respect to uniform convergence on \mathcal{C}* if there exists a probability space (Ω, \Pr) carrying a process W_P and processes S_λ , $0 < \lambda < \infty$, such that for

each λ the process S_λ has the same law (as a process on \mathcal{C}) as $(T_{\lambda P} - \lambda P)/\lambda^{1/2}$, and such that

$$\limsup_{\lambda \rightarrow \infty} \sup_{C \in \mathcal{C}} |(S_\lambda - W_P)(C)| = 0 \text{ a.s.}$$

The following result was proposed by E. B. Dynkin in a discussion in Oberwolfach, March 1979.

4.1. THEOREM. *For any probability measure P and P -Donsker class \mathcal{C} , $(T_{\lambda P} - \lambda P)/\lambda^{1/2}$ converges in law to W_P with respect to uniform convergence on \mathcal{C} .*

Proof. Take X_1, X_2, \dots , independent with distribution P . For each λ , $0 < \lambda < \infty$, let $n = n(\omega, \lambda)$ be a Poisson variable with parameter λ , independent of the X_i . Then we can write $T_{\lambda P} = n(\omega, \lambda) P_{n(\omega, \lambda)}$ (in law).

Now $(n(\omega, \lambda) - \lambda)/\lambda^{1/2}$ converges in law to a standard Gaussian variable as $\lambda \rightarrow \infty$. To replace this convergence by almost sure convergence of real random variables, we use the following standard procedure. For any probability distribution function F on \mathbb{R} and for $0 < y < 1$, let

$$F^{-1}(y) := \inf \{x: F(x) \geq y\}.$$

Suppose laws μ_m on \mathbb{R} with distribution functions F_m converge to a law μ_0 . Then $F_m^{-1}(y) \rightarrow F_0^{-1}(y)$ whenever the interval $F_0^{-1}\{y\}$ contains at most one point. Thus $F_m^{-1}(y) \rightarrow F_0^{-1}(y)$ for all y , $0 < y < 1$, if F_0 has an everywhere positive density, e.g. if it is a non-degenerate normal distribution function. Thus if $\mu_\lambda \rightarrow \mu_0$ as $\lambda \rightarrow \infty$, where μ_λ has distribution function F_λ and μ_0 has an everywhere strictly positive density, then, for $0 < y < 1$, $F_\lambda^{-1}(y) \rightarrow F_0^{-1}(y)$ as $\lambda \rightarrow \infty$ (continuously).

Now, taking a new probability space if necessary, we may assume that, for all ω ,

$$\lim_{\lambda \rightarrow \infty} (n(\omega, \lambda) - \lambda)/\lambda^{1/2} = H,$$

where H is a standard normal variable. Also, by 1.1, we can take $n^{1/2}(P_n - P) := v_n \rightarrow G_P$ uniformly on \mathcal{C} almost surely as $n \rightarrow \infty$, where the $n(\omega, \lambda)$ and H are independent of P_n and G_P .

Now $(n(\omega, \lambda) - \lambda)/\lambda \rightarrow 0$ a.s., so $n(\omega, \lambda)/\lambda \rightarrow 1$ a.s. and $n(\omega, \lambda) \rightarrow \infty$ a.s. Thus $v_{n(\omega, \lambda)} \rightarrow G_P$ uniformly on \mathcal{C} almost surely. So

$$\begin{aligned} & (n(\omega, \lambda) P_{n(\omega, \lambda)} - \lambda P)/\lambda^{1/2} \\ &= (n(\omega, \lambda)/\lambda)^{1/2} v_{n(\omega, \lambda)} + (n(\omega, \lambda) - \lambda) P/\lambda^{1/2} \rightarrow G_P + HP = W_P \end{aligned}$$

uniformly on \mathcal{C} almost surely as $\lambda \rightarrow \infty$, which completes the proof.

Added in proof. Theorem 4.1 extends a result known in the classical one-dimensional case: cf. P. Gaenssler and W. Stute, *Ann. Probability* 7 (1979), p. 193-243, Theorem 2.6.2 on p. 230-231.

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