

On the modulus of continuity of sample functions of Gaussian processes

By

Norio KONO

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1. Introduction

It is interesting that many important properties of sample functions of *G.p.*¹⁾ hold with probability 0 or 1. For some class of *G.p.*'s the modulus of continuity of sample functions is one of such properties. Let $\{X(s); s \in D\}$ be a real valued *G.p.* with a parameter space D . We shall throughout this paper assume the following.

(A.1) D is a compact convex subset of N -dimensional Euclidean space containing an open set with the usual Euclidean metric $\|s-t\|^2 = \sum_{i=1}^N (s_i - t_i)^2$.

(A.2) $\{X(s); s \in D\}$ has the mean $E[X(s)] = 0$ and with stationary increment $\sqrt{E[(X(s) - X(t))^2]} = \sigma(\|s-t\|)$, where $\sigma^2(x)$ ($\sigma(x) \equiv 0$) is concave near the origin and $\sigma(x)$ is a non-decreasing continuous function that satisfies

$$(1.1) \quad \int_0^{\infty} \sigma(e^{-x^2}) dx < +\infty.$$

This condition guarantees that the *G.p.* has continuous sample functions by modification due to the theorem of X. Fernique [5]. So we shall assume that the *G.p.* has continuous sample functions.

Now we are led to introduce the concept of upper class and lower class with respect to the modulus of uniform continuity or local continuity of sample functions.

Let $\varphi(x)$ be a non-increasing continuous function satisfying

1) *G.p.* means Gaussian process.

$\lim_{t \rightarrow 0} \varphi(x) = +\infty$. Then, after P. Lévy [10], we shall say that $\varphi(x)$ belongs to the *upper class* $\mathcal{U}^m(X)$ with respect to the modulus of uniform continuity if there exists a positive constant $\delta(\omega)$, with probability 1, such that $|X(s) - X(t)| < \sigma(\|s - t\|)\varphi(\|s - t\|)$ holds for any s, t with $\|s - t\| < \delta(\omega)$. We shall also say that $\varphi(x)$ belongs to the *lower class* $\mathcal{L}^m(X)$ with respect to the modulus of uniform continuity if there exists no such positive constant with probability 1. In the same manner we can define the *upper class* $\mathcal{U}^l(X)$ and the *lower class* $\mathcal{L}^l(X)$ with respect to the modulus of local continuity at any fixed point.

Investigations of the Hölder continuity of sample functions are formulated from our standpoint as follows: Sample functions of a *G.p.* are said to be uniform Hölder continuous of modulus $\varphi(x)$ if and only if $c\varphi(x) \in \mathcal{U}^m(X)$ or $\in \mathcal{L}^m(X)$ according as $c > 1$ or $0 < c < 1$ respectively. In a similar manner we can define the local Hölder continuity of modulus $\varphi(x)$. For some class of *G.p.*'s, however, we can only know properties that are weaker than the Hölder continuity in our sense, that is $c\varphi(x) \in \mathcal{U}^m(X)$ or $\mathcal{L}^m(X)$ ($\mathcal{U}^l(X)$ or $\mathcal{L}^l(X)$) provided $c > c_1$ or $c < c_2$ ($c_1 > c_2$) respectively, but we emphasize that c_1 and c_2 must coincide from Kolmogorov's 0-1 law as is explained in Section 6.

In case of Brownian motion, I. Petrovsky [14] proved the deciding condition that determines whether $\varphi(x)$ is of the class $\mathcal{U}^l(B)$ or $\mathcal{L}^l(B)$ (Kolmogorov test). This alternative corresponds to the regular points for the boundary value problem of the heat equation. K. L. Chung-P. Erdős-T. Sirao [3] proved an analogous deciding condition for uniform continuity. T. Sirao [17] showed deciding conditions for the case of Brownian motion with multi-dimensional parameter.

In this paper we shall give each deciding condition which determines $\mathcal{U}^m(X)$ or $\mathcal{L}^m(X)$ and $\mathcal{U}^l(X)$ or $\mathcal{L}^l(X)$ for a class of *G.p.*'s characterized by the conditions on the increment variance $\sigma(x)$. That is, Theorems 3 and 7 state that if $\sigma(x)$ is a nearly regular varying function (*n.r.v.f.* Definition 2 in Section 2), then $\varphi(x)$ belongs to $\mathcal{U}^m(X)$ or $\mathcal{L}^m(X)$ according as

$$(1.2) \quad I_u(\sigma, \varphi) = \int_{+0} x^{N-1} \exp \left\{ -\frac{1}{2} \varphi^2(x) \right\} \left[\sigma^{-1} \left(\frac{\sigma(x)}{\sqrt{\log 1/x}} \right) \right]^{2N} \varphi(x) dx$$

converges or diverges respectively. Also $\varphi(x)$ belongs to $\mathcal{U}(X)$ or $\mathcal{L}(X)$ according as

$$(1.3) \quad I_l(\sigma, \varphi) = \int_{+0} x^{N-1} \exp \left\{ -\frac{1}{2} \varphi^2(x) \right\} \left[\sigma^{-1} \left(\frac{\sigma(x)}{\sqrt{\log_{(2)} 1/x}} \right) \right]^{2N} \varphi(x) dx$$

converges or diverges respectively.

Recently, T. Sirao-H. Watanabe [18] have treated the *G.p.*'s with stationary increments on an interval for which $\sigma^2(x)$ satisfies the inequality

$$cx^\alpha (\log 1/x)^\beta \leq \sigma^2(x) \leq c'x^\alpha (\log 1/x)^\beta$$

and is also concave near the origin. Their results have shown that the upper class and the lower class depend on α but not on β , that is, both classes depend only on the exponent of *n.r.v.f.* in this case. Here arises the question of whether or not the upper and lower classes are the same for any two *G.p.*'s with increment variances which are *n.r.v.f.* with the same exponent. Our theorems, however, state that both classes can be different even if corresponding $\sigma(x)$'s are *n.r.v.f.* with the same exponent. For example, let $\{X_i(t); 0 \leq t \leq 1\}$ ($i=1, 2, 3$) be three *G.p.*'s with stationary increments which have increment variances $\sigma_i(x)$ such that

$$\begin{aligned} \sigma_1(x) &\asymp x^{\alpha \beta}, \\ \sigma_2(x) &\asymp x^\alpha \exp \{ (\log 1/x)(\log_{(2)} 1/x)^{-\beta} \}, \\ \sigma_3(x) &\asymp x^\alpha \exp \{ (\log 1/x)(\log_{(3)} 1/x)^{-\gamma} \}, \\ &\quad (0 < \alpha \leq 1/2, \beta, \gamma > 0). \end{aligned}$$

1) By concavity and monotonicity of $\sigma^2(x)$, $\sigma(x)$ has the inverse function near the origin. The integral \int_{+0}^x means $\lim_{\epsilon \rightarrow 0} \int_{\epsilon}^x$ if $\sigma^2(x)$ is concave at $(0, u]$.

2) $\log_{(2)} x = \log \log x$, $\log_{(n)} x = \log \log_{(n-1)} x$.

3) $f(x) \asymp g(x)$ means $0 < \liminf f(x)/g(x) \leq \limsup f(x)/g(x) < \infty$.

Then we have

$$(1.4) \quad \mathcal{U}^\beta(X_1) = \mathcal{U}^\beta(X_2) \quad \text{and} \quad \mathcal{L}^\beta(X_1) = \mathcal{L}^\beta(X_2)$$

when $\beta \geq 1$, but these equalities never hold when $0 < \beta < 1$, and similarly we get

$$(1.5) \quad \mathcal{U}^\gamma(X_1) = \mathcal{U}^\gamma(X_2) \quad \text{and} \quad \mathcal{L}^\gamma(X_1) = \mathcal{L}^\gamma(X_2)$$

when $\gamma \geq 1$, but these equalities never hold when $0 < \gamma < 1$.

T. Kawada [8] proved an analogous deciding condition in case of *G.p.* with increment variance $E[(X(s) - X(t))^2] = \|s - t\|^\alpha$ ($0 < \alpha \leq 1$) over N -dimensional Euclidean space. Many authors have investigated the Hölder continuity of sample functions [1], [13], [19].

We shall give in Theorems 1 and 2 the uniform Hölder continuity obtained by M. B. Marcus in case of a *G.p.* on an interval. It is remarkable, as is explained in Remarks 1 and 2, that if $\sigma(x)$ is a *n.r.v.f.* the modulus of uniform Hölder continuity equals $\sqrt{2 \log N_\varepsilon(D)}$, where $N_\varepsilon(D)$ is the minimal number of ε -covering of D . This fact is valid even if D is some infinite dimensional compact subset of a Hilbert space, (which contains the results of P. T. Strait [19]).

We shall also give some results on the modulus of local continuity of sample functions (Section 5).

Our theorems suggest to us that it is useful to investigate the modulus of continuity of sample functions of *G.p.*'s under some classification. Now we shall propose the following classification of *G.p.*'s. Let $\{X(s)\}$ be a *G.p.* satisfying our assumptions. Set

$$(1.6) \quad \sigma(x) = F_\sigma(x) / \sigma(x),$$

where $F_\sigma(x) = \int_0^\infty \sigma(xe^{-u^2}) du$.

Class I. $\{X(s)\}$ belongs to Class I if $\sigma(x)$ is bounded.

Class II. $\{X(s)\}$ belongs to Class II if $\sigma(x)$ is not bounded but satisfies

$$(1.7) \quad \lim_{x \rightarrow 0} \sigma(x) / \sqrt{\log 1/x} = 0.$$

Class III. $\{X(s)\}$ belongs to Class III if $\sigma(x)$ satisfies

$$(1.8) \quad \sigma(x) \asymp \sqrt{\log 1/x}.$$

Class IV. The rest of the *G.p.*'s.

We shall briefly discuss in Section 3 the meaning of our classification and difference of classes. Also we shall give a classification if the function $\sigma(x)$ is a *n.r.v.f.* or a *n.s.v.f.* (Definition 4 in Section 2).

The content of each section is as follows.

1. Introduction.
2. Some notation and some definitions.
3. About our classification.
4. Theorems on the modulus of uniform continuity.
5. Theorems on the modulus of local continuity.
6. Preliminary lemmas.
7. Proofs of Theorems 1 and 2.
8. Proofs of Theorems 3 and 4.
9. Proofs of Theorems 5 and 6.
10. Proofs of Theorems 7 and 8.
11. Proofs of Propositions 1~4.

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2. Some notation and some definitions

First we shall give the following modified definitions of regular varying function with exponent $\alpha > 0$, nearly regular varying function, slowly varying function, and nearly slowly varying function due to J. Karamata [7].

Definition 1. Let $r(x)$ be a positive continuous function defined on a semi-closed interval $(0, u_0]$. Then $r(x)$ is called a regular varying function with exponent $\alpha (> 0)$ (*r.v.f.* (α)) if it holds $\lim_{t \rightarrow 0} r(tx)/r(x) = t^\alpha$ for any $t > 0$.

Definition 2. We shall say that $f(x)$ is a nearly regular varying function with exponent α (*n.r.v.f.* (α)) if there exists a *r.v.f.* (α) $r(x)$ such that $r(x) \asymp f(x)$.

Definition 3. Let $s(x)$ be a function defined in Definition 1. Then $s(x)$ is called a slowly varying function (*s.v.f.*) if it holds $\lim_{x \rightarrow \infty} s(tx)/s(x) = 1$ for any $t > 0$.

Definition 4. We shall say that $g(x)$ is a nearly slowly varying function (*n.s.v.f.*) if there exists a *s.v.f.* $s(x)$ such that $s(x) \asymp g(x)$.

It is well known (for instance [20]) that $r(x)$ is a *r.v.f.* (α) if and only if there exists a *s.v.f.* $s(x)$ such that $r(x) = x^\alpha s(x)$ and $s(x)$ is expressible as follows:

$$(2.1) \quad s(x) = b(x) \exp \left\{ - \int_x^c \frac{a(u)}{u} du \right\} \quad (c \in (0, u_0)),$$

where $a(x)$ and $b(x)$ are continuous functions such that $\lim_{x \rightarrow \infty} a(x) = 0$ and $\lim_{x \rightarrow \infty} b(x) > 0$.

Definition 5. We shall say the function $a(x)$ of (2.1) a *structure function* of the *s.v.f.* or the *n.s.v.f.*.

The following definition is useful to describe our assertions.

Definition 6. For two functions $f(x)$ and $g(x)$ defined near the origin, we shall denote by $f \gg g$ ($x \downarrow 0$) if there exists $u_0 > 0$ such that $f(x) \geq g(x)$ holds for any $0 < x < u_0$. We shall say that $f_\lambda \gg g_\lambda$ ($x \downarrow 0$) holds *uniformly* if there exists a constant $u_0 > 0$ independent of λ such that $f_\lambda(x) \geq g_\lambda(x)$ holds for any $0 < x < u_0$. By an analogous way we shall define $a_n \gg b_n$ ($n \rightarrow \infty$) for two sequences $\{a_n\}_{n=1}^\infty$ and $\{b_n\}_{n=1}^\infty$.

Next we shall prepare some notation which are used repeatedly.

$$\Phi(a) = \frac{1}{\sqrt{2\pi}} \int_a^\infty e^{-u^2/2} du,$$

$$\Phi_r(a, b) = \int_a^b \int_b^{\sqrt{2\pi}} \frac{1}{\sqrt{2\pi} \sqrt{1-r^2}} \exp \left\{ - \frac{x^2 - 2r \frac{xy+y^2}{\sqrt{1-r^2}}}{2(1-r^2)} \right\} dx dy.$$

$$F_{\sigma}(x) = \int_0^x \sigma(xe^{-u^2}) du,$$

$$G_{\sigma}(x) = \int_x^{\infty} \sigma(e^{-u^2}) du,$$

$$a \vee b = \max(a, b), \quad a \wedge b = \min(a, b).$$

Let (S, ρ) be a compact metric space, and K is a compact subset of S with the diameter $d(K)$. Then we define

$$N_{\varepsilon}(K) = \inf \# \left\{ K \subset \bigcup_{i=1}^l A_i \quad \text{such that } d(A_i) \leq 2\varepsilon \right\},$$

$$M_{\varepsilon}(K) = \sup \# \left\{ (t_1, t_2, \dots, t_n) \subset K \quad \text{such that } \rho(t_i, t_j) > \varepsilon \right. \\ \left. \text{for any } i \neq j \right\}.$$

We shall denote by $c_i(\varepsilon)$ a positive constant dependent on ε and denote by d_i an absolute constant.

3. About our classification

First we shall give a classification if the function $\sigma(x)$ is a *n.r.v.f.* or a *n.s.v.f.*.

Proposition 1. *If $\sigma(x)$ is a n.r.v.f., then the process is of Class I, and if $\sigma(x)$ is a n.s.v.f., it is not of Class I.*

Proposition 2. *Assume that $\sigma(x)$ is a n.s.v.f. with a structure function $a(x)$ which satisfies the following two conditions:*

$$(i) \quad a(x) \gg \frac{\delta}{\log 1/x} \quad (x \downarrow 0)$$

holds for any $\delta > 0$.

(ii) *There exists a constant $c_3 > 0$ such that*

$$a(x^{1-\varepsilon}) \gg c_3 a(x) \quad (x \downarrow 0)$$

uniformly for any $0 \leq \varepsilon \leq 1$. Then the process is of Class II.

Proposition 3. *If $\sigma(x)$ is a n.s.v.f. with a structure function $a(x)$ such that*

1) $\#l$ denotes the cardinal number of l .

$$\frac{c_1}{\log 1/x} \gg a(x) \gg \frac{c_2}{\log 1/x} \quad (x \downarrow 0), (c_1 > c_2 > 1/2),$$

then the process is of Class III.

Proposition 4. Assume that $\sigma(x)$ is a n.s.v.f. with a structure function $a(x)$ such that

$$\begin{aligned} & \frac{1}{2 \log 1/x} + \frac{1}{(\log 1/x)(\log_{(2)} 1/x)} + \cdots + \frac{1}{(\log 1/x) \cdots (\log_{(m-1)} 1/x)} \\ & + \frac{c_6}{(\log 1/x) \cdots (\log_{(m)} 1/x)} \gg a(x) \gg \frac{1}{2 \log 1/x} + \frac{1}{(\log 1/x)(\log_{(2)} 1/x)} \\ & + \cdots + \frac{1}{(\log 1/x) \cdots (\log_{(m-1)} 1/x)} + \frac{c_7}{(\log 1/x) \cdots (\log_{(m)} 1/x)}, \end{aligned}$$

(x ↓ 0), (c₆ > c₇ > 1).

Then the process is of Class IV.

These propositions are proved in Section 11. Here we shall give some examples.

Class	$\sigma(x) \asymp$
I	n.r.v.f., $x^\alpha (\log 1/x)^\beta$ ($0 < \alpha$, $-\infty < \beta < +\infty$), $x^\alpha \exp \{ \pm (\log 1/x)^\beta \}$ ($0 < \alpha$, $-\infty < \beta < 1$),
II	$\exp \{ -(\log 1/x)^\beta (\log_{(2)} 1/x)^\gamma \}$ ($0 < \beta < 1$, $-\infty < \gamma < +\infty$),
III	$(\log 1/x)^{-\gamma} (\log_{(2)} 1/x)^\delta$ ($\gamma > 1/2$, $-\infty < \delta < +\infty$),
IV	$(\log 1/x)^{-\nu_1} (\log_{(2)} 1/x)^{-\nu_2} \cdots (\log_{(m)} 1/x)^{-\nu_m} (\log_{(m+1)} 1/x)^{\delta_1}$ $\cdots (\log_{(m+n)} 1/x)^{\delta_n}$, ($\delta_i > 1$, $-\infty < \delta_i < +\infty$, $i=2, \dots, n$).

Now we shall briefly discuss the meaning of our classification and the difference of classes.

In case of G.p.'s with stationary increments on an interval, it follows from Theorem 1 and Theorem 2 that if a G.p. is of Class I or II the sample functions are uniform Hölder continuous of modulus $\sqrt{2} \log 1/x$, while if the process is of Class III the sample functions are uniformly Hölder continuous of modulus $\sqrt{c \log 1/x}$ which is of $\mathcal{U}^c(X)$ or $\mathcal{L}^c(X)$ according as $c > c_s \geq 2$ or

$c < 2$ respectively. If, in particular, we assume that $\sigma(x)$ is a *n.r.v.f.* the process belongs to Class I and we have the conditions that decide whether $\varphi(x)$ is of $\mathcal{U}''(X)$ or $\mathcal{L}''(X)$ and $\mathcal{U}'(X)$ or $\mathcal{L}'(X)$ as described already.

The crucial difference between Class I and Class II may be seen by a particular choice of $\varphi(x)$ as follows. If $\sigma(x)$ is a *n.r.v.f.* (i.e. the *G.p.* is of Class I) we see that $\varphi(x) = \sqrt{2 \log 1/x + c \log_{c_2} 1/x}$ belongs to $\mathcal{U}''(X)$ or $\mathcal{L}''(X)$ depending on the constant c . While if $\sigma(x)$ is a *n.s.v.f.* with a structure function $a(x)$ the process is of Class II only under certain conditions on $a(x)$ and $\varphi(x) = \sqrt{2 \log 1/x + ca(x)^{-1} \log_{c_2} 1/x} \in \mathcal{U}''(X)$ or $\in \mathcal{L}''(X)$ (Theorem 4). Deciding conditions for $\mathcal{U}''(X)$ or $\mathcal{L}''(X)$ ($\mathcal{U}'(X)$ or $\mathcal{L}'(X)$) of Class II~IV are unknown.

4. Theorems on the modulus of uniform continuity

First we shall give the uniform upper bound of sample functions.

Theorem 1. *Let $\{X(s) : s \in D\}$ be a *G.p.* satisfying our assumptions (A. 1), (A. 2) and assume that $\sigma(x) \sqrt{\log 1/x}$ is a non-decreasing function near the origin. Then*

$$(4.1) \quad \overline{\lim}_{|s-t| \rightarrow 0} \frac{|X(s) - X(t)|}{\sigma(\sqrt{|s-t|}) \varphi_\varepsilon(\sqrt{|s-t|})} \leq 1$$

holds with probability one for any $\varepsilon > 0$, where

$$(4.2) \quad \varphi_\varepsilon(x) = \sqrt{(2 + \varepsilon)N \log 1/x + c_\varepsilon(\varepsilon)G_\varepsilon(\sqrt{N \log 1/x})/\sigma(x)}.$$

Remark 1. Theorem 1 is still valid under the non-decreasingness of $\sigma(x) \sqrt{H(\log 1/x)}$ and the following conditions (A. 1)^(*), (A. 2)^(*) instead of (A. 1), (A. 2) by substituting

$$(4.3) \quad \varphi_\varepsilon(x) = \sqrt{(4 + \varepsilon)H(\log 1/x) + c_{\varepsilon,0}(\varepsilon)G_\varepsilon''(\sqrt{H(\log 1/x)})/\sigma(x)}$$

where $G_\varepsilon''(x) = \int_x^\infty \sigma(\exp\{-H^{-1}(u^2)\}) du$.

(A. 1)^(*) The parameter space (D, ρ) is a compact centered metric

1) For any two points $s_1, s_2 \in D$ there exists $s_0 \in D$ such that $\rho(s_0, s_1) = \rho(s_0, s_2) = 1/2 \rho(s_1, s_2)$.

space¹⁾ satisfying

$$\exp \{(1+\delta)H(\log 1/\varepsilon)\} \gg N_\varepsilon(D) \gg \exp \{(1-\delta)H(\log 1/\varepsilon)\} \quad (\varepsilon \downarrow 0)$$

for any $\delta > 0$, where $1/H(1/x)$ is a strictly increasing *r.v.f.* ($\mu \geq 1$).

$$(A. 2)^{(**)} \quad G_\sigma^H(0) < +\infty.$$

Furthermore assume that (D, ρ) satisfies

$$(A. 1)^{(***)} \quad \lim_{\varepsilon \downarrow 0} \frac{N_\varepsilon(B(t, a\varepsilon))}{N_\varepsilon(D)} = 0, \quad (B(t, r) = \{s \in D; \|t-s\| \leq r\}),$$

uniformly in $t \in D$ for any $a > 1$. Then we can choose $\varphi_\varepsilon(x)$ instead of (4.3) as follows:

$$(4.4) \quad \varphi_\varepsilon(x) = \sqrt{(2+\varepsilon)H(\log 1/x)} + c_{11}(\varepsilon)G_\sigma^H(\sqrt{H(\log 1/x)})/\sigma(x).$$

Corollary 1. *When a G.p. $\{X(s); s \in D\}$ belongs to Class I, II or III and satisfies the assumption of Theorem 1, we have*

$$(4.5) \quad \overline{\lim}_{\|s-t\| \rightarrow 0} \frac{|X(s) - X(t)|}{\sigma(\|s-t\|) \sqrt{2N \log 1/\|s-t\|}} \leq c_{12}$$

with probability one. Furthermore if the process belongs to Class I or II we can choose $c_{12} = 1$, but if the process belongs to Class III it is an open problem whether $c_{12} = 1$ or not.

Corollary 2. *If a G.p. $\{X(s); s \in D\}$ satisfies the conditions of Proposition 4, we have*

$$(4.6) \quad \overline{\lim}_{\|s-t\| \rightarrow 0} \frac{|X(s) - X(t)|}{\sigma(\|s-t\|) \sqrt{2N \log 1/\|s-t\|} (\log_{(2)} 1/\|s-t\|) \cdots (\log_{(m)} 1/\|s-t\|)} \leq c_{13} < +\infty$$

with probability one.

Next we shall give the uniform lower bound of sample functions.

Theorem 2. *Let $\{X(s); s \in D\}$ be a G.p. satisfying our assumptions (A.1) and (A.2). Then we have*

$$(4.7) \quad \overline{\lim}_{\|s-t\| \rightarrow 0} \frac{|X(s) - X(t)|}{\sigma(\|s-t\|) \sqrt{2N \log 1/\|s-t\|}} \geq c_{14} \geq 1/\sqrt{2}$$

with probability one. Furthermore if we assume that $\sigma(x)$ is a n.r.v.f. we can set $c_{14}=1$.

Remark 2. Theorem 2 can be extended as follows. If the parameter space (D, ρ) satisfies (A. 1)^(*) and have the four point property¹⁾ we have

$$(4. 8) \quad \overline{\lim}_{n \rightarrow \infty} \frac{1}{n} \frac{\sum_{i=1}^n |X(s_i) - X(t_i)|}{\sqrt{2H(\log 1/||s-t||)}} \geq c_{15} \geq 1/\sqrt{2}$$

with probability one. Furthermore if we assume that $\sigma(x)$ is a n.r.v.f. we can set $c_{15}=1$.

For example define $D = \{(a_1, a_2, \dots) \in l^2 : \sum_{i=1}^n a_i^2 \leq 2^{-n} \text{ for any } n\}$ with usual l^2 -metric. Then $N_*(D)$ satisfies (A. 1)^(*) for $H(x) = x^2/2$ ([11]).

Remark 3. If $D=[0, 1]$, we can set $c_{14}=1$ ([13]).

Now we shall give a deciding condition for uniform upper class or lower class.

Theorem 3. Let $\{X(s) : s \in D\}$ be a G.p. satisfying our assumptions (A. 1) and (A. 2). Furthermore we assume that $\sigma(x)$ is a n.r.v.f.. Then we have

$$(4. 9) \quad \varphi \in \mathcal{U}^u(X) \quad \text{if} \quad I_n(\sigma, \varphi) < +\infty$$

and

$$\varphi \in \mathcal{L}^u(X) \quad \text{if} \quad I_n(\sigma, \varphi) = -\infty.$$

As the corollary of Theorem 3, we shall give a condition of invariance of $\mathcal{U}^u(X)$ and $\mathcal{L}^u(X)$ which we have pointed out in Introduction.

Let $\{X_i(s) : s \in D\}$ ($i=1, 2$) be two G.p.'s satisfying our assumptions (A. 1) and (A. 2). Set $\sigma_\alpha(x) \asymp x^\alpha$ ($0 < \alpha \leq 1/2$) and $\sigma^\tau(x) \asymp x^\alpha \tau(x)$, where $\tau(x)$ is a s.v.f. with a structure function $a(x)$. Then there exists a s.v.f. $\hat{\tau}(x)$ such that

$$(4. 10) \quad \sigma_2^{-1} \left(\frac{\sigma_1(x)}{\sqrt{\log 1/x}} \right) \asymp x(\log 1/x)^{-1/2} \hat{\tau}(x).$$

1) Any four points of D can be imbeded in 3-dimensional Euclidean space without changing their metrics each other.

Corollary 3. *If*

$$(4.11) \quad \phi(x) \asymp c_{16},$$

then we have

$$(4.12) \quad \mathcal{U}^*(X_1) = \mathcal{U}^*(X_2) \quad \text{and} \quad \mathcal{L}^*(X_1) = \mathcal{L}^*(X_2).$$

In particular,

$$(4.13) \quad |a(x) \log_{(2)} 1/x| \leq c_{17}$$

guarantees the relation (4.11).

Corollary 4. *On the other hand, (4.12) does not hold if $a(x)$ satisfies the condition (ii) of Proposition 2 and*

$$(4.14) \quad |a(x) \log_{(2)} 1/x| \gg c_{18} \log_{(2)} 1/x \quad (x \downarrow 0).$$

In case of $\sigma(x)$ being not a n.r.v.f. we have only restricted results as follows.

Theorem 4. *If $\sigma(x)$ is a n.s.v.f. with a structure function $a(x)$ which satisfies the following conditions:*

(i) *$a(x)$ is non-decreasing and there exists a constant $c_{19}(\delta)$ satisfying $\lim_{\delta \downarrow 0} c_{19}(\delta) = 1$ such that*

$$a(x^{1-\delta}) \gg c_{19}(\delta) a(x) \quad (x \downarrow 0)$$

holds uniformly.

(ii) *There exists a constant $1 > \beta > 0$ such that*

$$a(x) \gg (\log 1/x)^{\beta-1} \quad (x \downarrow 0).$$

Then

$$(4.15) \quad \varphi_\varepsilon(x) = \sqrt{2N \log 1/x + 2N(1+\varepsilon)(\log_{(2)} 1/x)} / a(x) \\ \in \mathcal{U}^*(X) \quad \text{if } \varepsilon > 0,$$

and

$$\in \mathcal{L}^*(X) \quad \text{if } \varepsilon < 0.$$

The conditions (i) and (ii) imply that this process is of Class II by Proposition 2.

5. Theorems on the modulus of local continuity

In this section we shall give some theorems corresponding to the case of uniform continuity. Theorem 5 and 6 are concerned with the local Hölder continuity. M. B. Marcus [13] has proved about this in case of $D=[0, 1]$. We shall adjust his results in case of $\sigma(x)$ satisfying some regularity conditions. Theorem 7 gives a deciding condition that determines whether $\varphi(x)$ belongs to $\mathcal{U}(X)$ or $\mathcal{L}(X)$.

Let $\{X(s) : s \in D\}$ be a G.p. with our assumptions (A. 1) and (A. 2).

Theorem 5. (i) *If $\sigma(x)$ is a n.r.v.f., then for any fixed point $s_0 \in D$.*

$$(5.1) \quad \overline{\lim}_{\|s-s_0\| \rightarrow 0} \frac{|X(s) - X(s_0)|}{\sqrt{\sigma(\|s-s_0\|)} \sqrt{2 \log_{(2)} 1/\|s-s_0\|}} \leq 1$$

holds with probability one.

(ii) *Assume that $\sigma(x)$ is a n.s.v.f. with a structure function $a(x)$ which satisfies following condition:*

a) *$a(x)$ is non-decreasing and satisfies*

$$a(x^{1+\varepsilon}) \gg c_{20} a(x) \quad \text{uniformly for } 0 \leq \varepsilon \leq 1.$$

b) $a(x) \gg \frac{\gamma}{\sqrt{\log 1/x}}, \quad (x \downarrow 0) \quad (\gamma > 1/2).$

Then

$$(5.2) \quad \overline{\lim}_{\|s-s_0\| \rightarrow 0} \frac{|X(s) - X(s_0)|}{\sqrt{\sigma(\|s-s_0\|)} \varphi(\|s-s_0\|)} \leq c_{21}$$

holds with probability one, where

$$(5.3) \quad \varphi(x) = \sqrt{\log_{(2)} 1/x} \sqrt{1/a(x)},$$

(iii) *If $\sigma(x)$ is n.s.v.f. with a non-decreasing structure function $a(x)$ satisfying the condition of Proposition 4, then*

$$(5.4) \quad \overline{\lim}_{\|s-s_0\| \rightarrow 0} \frac{|X(s) - X(s_0)|}{\sqrt{\sigma(\|s-s_0\|)} \sqrt{\log \bar{1}/\|s-s_0\|} \sqrt{\log_{(2)} 1/\|s-s_0\|} \cdots \sqrt{\log_{(m)} 1/\|s-s_0\|}} \leq c_{22}$$

holds with probability one.

Next we shall give the local lower bound of sample functions of a $G.p.$.

Theorem 6. (i) *If $\sigma(x)$ is a n.s.v.f.,*

$$(5.6) \quad \overline{\lim}_{\|s-s_0\| \rightarrow 0} \frac{|X(s) - X(s_0)|}{\sigma(\|s-s_0\|) \sqrt{2 \log_{(2)} 1/\|s-s_0\|}} \geq 1$$

holds with probability one.

(ii) *Assume that $\sigma(x)$ is a n.s.v.f. of a structure function $a(x)$ satisfying the following condition:*

$$(5.7) \quad 0 \leq a(x+h) - a(x) \ll c_2 h a(x)/x \quad (x \downarrow 0),$$

holds uniformly for $h > 0$. Then

$$(5.8) \quad \overline{\lim}_{\|s-s_0\| \rightarrow 0} \frac{|X(s) - X(s_0)|}{\sigma(\|s-s_0\|) \varphi(\|s-s_0\|)} \geq c_2 > 0$$

holds with probability one, where $\varphi(x)$ is given by (5.3).

Now we shall give a deciding condition which determines $\mathcal{U}'(X)$ or $\mathcal{L}'(X)$ for the same class of $G.p.$'s as Theorem 3. This theorem includes the cases of Theorem 5, (i) and Theorem 6, (i).

Theorem 7. *Under the same conditions of Theorem 3, we have*

$$(5.9) \quad \varphi \in \mathcal{U}'(X) \quad \text{if} \quad I_1(\sigma, \varphi) < +\infty,$$

$$\text{and} \quad \varphi \in \mathcal{L}'(X) \quad \text{if} \quad I_1(\sigma, \varphi) = +\infty.$$

We have the following corollaries of Theorem 7 by the same way as Corollaries 3 and 4.

Corollary 5. *Under the same situation of Corollary 3, we have*

$$(5.10) \quad \sigma_x^{-1} \left(\frac{\sigma_2(x)}{\sqrt{\log_{(2)} 1/x}} \right) \asymp x (\log_{(2)} 1/x)^{-1/2} \tau(x),$$

where $\tau(x)$ is a s.v.f. and if

$$(5.11) \quad \tau(x) \asymp c_2.$$

we get

$$(5.12) \quad \mathcal{U}'(X_1) = \mathcal{U}'(X_2) \quad \text{and} \quad \mathcal{L}'(X_1) = \mathcal{L}'(X_2).$$

In particular,

$$(5.13) \quad |a(x)\log_{c_3} 1/x| \leq c_2$$

guarantees the relation (5.12).

Corollary 6. *On the other hand, (5.12) does not hold if $a(x)$ satisfies the condition (ii) of Proposition 2 and*

$$(5.14) \quad |a(x)\log_{c_3} 1/x| \gg c_{26} \log_{c_4} 1/x \quad (x \downarrow 0).$$

In case $\sigma(x)$ being not a *n.r.v.f.* we have only restricted results as follows.

Theorem 8. *Assume that $\sigma(x)$ is a *n.s.v.f.* with a structure function $a(x)$ which satisfies the condition (i) of Theorem 4 and a following condition:*

(i) *There exists a positive constant $1 > \delta > 0$ such that*

$$a(x) \gg (\log_{c_2} 1/x)^{\delta-1} \quad (x \downarrow 0).$$

Then

$$(5.15) \quad \varphi_\varepsilon(x) = \sqrt{2 \log_{c_2} 1/x + N(1+\varepsilon)(\log_{c_3} 1/x)/a(x)} \\ \in \mathcal{U}^\varepsilon(X) \quad \text{if } \varepsilon > 0$$

and

$$\in \mathcal{L}^\varepsilon(X) \quad \text{if } \varepsilon < 0.$$

6. Preliminary lemmas

Before we prove the theorems, we shall prepare some useful lemmas.

Lemma 1 (0-1 law). *Let $\varphi(x)$ be a continuous function satisfying $\lim_{x \downarrow 0} \varphi(x) = +\infty$. Then the random variables*

$$(6.1) \quad \lim_{\sigma \rightarrow 0} \frac{|X(s) - X(t)|}{\sigma(\|s - t\|) \varphi(\|s - t\|)}$$

and

$$(6.2) \quad \lim_{\sigma \rightarrow 0} \frac{|X(s) - X(s_0)|}{\sigma(\|s - s_0\|) \varphi(\|s - s_0\|)} \quad \text{for a fixed } s_0,$$

are constants with probability one.

The proof of Lemma 1 is based on the fact that $\{X(s)\}$ is expanded uniformly in s with probability one in the form $\sum_{n=1}^{\infty} \sqrt{\lambda_n} \varphi_n(s) X_n(\omega)$, where $\{\lambda_n\}$, $\{\varphi_n(s)\}$ are eigen values and normalized eigen functions respectively of following integral equation,

$$(6.3) \quad \lambda \varphi(s) = \int_D E[X(s)X(t)] \varphi(t) dt. \quad (\text{see [6], [21]}).$$

By the definition of upper class or lower class and $I_u(\sigma, \varphi)$ or $I_l(\sigma, \varphi)$ we have

Lemma 2. (i) If $\varphi_1 \in \mathcal{U}^*(X)(\mathcal{U}^l(X))$ and $\varphi_2 \gg \varphi_1$, then $\varphi_2 \in \mathcal{U}^*(X)(\mathcal{U}^l(X))$.

(ii) $\varphi_3 \in \mathcal{L}^*(X)(\mathcal{L}^l(X))$ and $\varphi_3 \gg \varphi_1$, then $\varphi_1 \in \mathcal{L}^*(X)(\mathcal{L}^l(X))$.

(iii) If $I_u(\sigma, \varphi) < +\infty$, we have $\varphi \gg \sqrt{N \log 1/x}$ and $I_u(\sigma, \varphi_3) < +\infty$ by setting $\varphi_3 = (\varphi \vee \sqrt{N \log 1/x}) \wedge \sqrt{3N \log 1/x}$.

(iv) If $I_u(\sigma, \varphi) = +\infty$, then $I_u(\sigma, \varphi_3) = +\infty$ and if $\varphi_3 \in \mathcal{L}^*(X)$ then $\varphi \in \mathcal{L}^*(X)$.

(v) The similar facts to (iii) and (iv) hold for $I_l(\sigma, \varphi)$.

The proof of Lemma 2 is in the same way as that of T. Sirao [17].

Lemma 3 (Borel-Cantelli). Let $\{A_n\}$ be a sequence of events.

(i) If $\sum_{n=1}^{\infty} P(A_n) < +\infty$, then $P(\overline{\lim}_{n \rightarrow \infty} A_n) = 0$.

(ii) a) $\sum_{n=1}^{\infty} P(A_n) = +\infty$,

b) For each n there exists finite subsequence $I_n = \{n \leq n_1 < n_2 < \dots < n_{l(n)}\}$ such that

(b. 1) $\sum_{i \in I_n} P(A_n \cap A_i) \leq d_1 P(A_n)$

(b. 2) $P(A_n \cap A_i) \leq d_2 P(A_n) P(A_i)$ if $i \in I_n, i \geq n$,

then $P(\overline{\lim}_{n \rightarrow \infty} A_n) \geq \frac{1}{d_2}$.

The proof is obvious from Shwarz' inequality (c.f. [13] Lemma 1).

Lemma 4. (i) [12, p. 4]

$$\frac{e^{-x^2/2}}{\sqrt{2\pi}(x+1/x)} \leq \Phi(x) \leq \frac{e^{-x^2/2}}{\sqrt{2\pi}x} \quad (x > 0).$$

(ii) [3, Lemma 3]

$$\Phi_r(a, b) \leq d_s \Phi(a) \Phi(b) \quad (a, b > 0)$$

for any $-1 < r < 1/ab$, where d_s is an absolute constant independent of a, b and r .

(iii) [3, Lemma 4]

$$\Phi_r(a, b) \leq d_s \exp\left\{-\frac{1-r}{4}b^2\right\} \Phi(a)$$

for any $b \geq a \geq 0$ and $1 > r \geq 0$.

(iv) [17, Lemma 2]

$$\int_{-\infty}^a \int_b^{\infty} \frac{1}{\sqrt{1-r^2}} \exp\left\{-\frac{x^2 - 2rxy + y^2}{2(1-r^2)}\right\} dx dy$$

is a non-increasing function of r for fixed a and b ($0 < a < b$).

(v) [2, p. 508] (Stepian)

Let $R = \{r_{i,j}\}$ be a $N \times N$ symmetric positive definite matrix with $r_{ii} = 1$ ($i = 1, 2, \dots, N$). Define

$$Q(c, R) = \int_{-\infty}^c \dots \int_{-\infty}^c \frac{1}{\sqrt{\det R}} \exp\left\{-\frac{1}{2}(x, R^{-1}x)\right\} dx_1 \dots dx_N.$$

Then $Q(c, R)$ is an increasing function of the arguments $r_{i,j}$.

Lemma 5. Let $s(x)$ be a s.v.f.. Then it follows by (2.1) that for any $\varepsilon > 0$ there exists a constant $c(\varepsilon) > 0$ such that

$$c(\varepsilon)t^{-\varepsilon} \leq \frac{s(tx)}{s(x)} \leq c(\varepsilon)t^\varepsilon$$

for any $t \geq 1, 0 < xt \leq d_s$.

Lemma 6. Let $f(x)$ be a concave function which is non-decreasing with $f(0) = 0$.

(i) $|f(x_1) + f(x_2) - f(x_3) - f(x_4)| \leq f(|x_1 + x_2 - x_3 - x_4|)$

(ii) $f(x_1) + f(x_2) - f(x_3) \geq f(x_1 + x_2 - x_3)$

if $x_1 + x_2 \geq x_3 \geq x_1 \vee x_2$.

Lemma 7. For any 4 points A, B, A' and B' of Euclidean space,

$$(i) \quad |\overline{AB'} + A'\overline{B} - \overline{AA'} - \overline{BB'}| \leq 4\overline{AB} \cdot \overline{A'B'} / R,$$

where $R = \max \{ \overline{AA'} + \overline{A'B}, \overline{AB'} + \overline{B'B}, \overline{A'A} + \overline{AB'}, \overline{B'B} + \overline{A'B} \}$

(ii) If $\overline{AB'} \geq \overline{AA'}$ and $A'B \geq \overline{BB'}$ then

$$\overline{AB'} + \overline{A'B} - \overline{AA'} - \overline{BB'} \leq 2\overline{AB} \cdot \overline{A'B'} / r$$

where $r = \min \{ \overline{AA'} + \overline{AB'}, A'B + \overline{BB'} \}$.

\overline{PQ} denotes the distance of P and Q .

Lemma 8 [9]. Let $D \subset R^N$ be a compact convex subset which contains an open subset. Then for any $B(t, r) = \{s \in D : \|s - t\| \leq r\}$, $t \in D$,

$$d_*(r/\varepsilon)^N \geq N_t(B(t, r)) \geq d_*(r/\varepsilon)^N,$$

where d_* and d_r are constants independent of r and ε .

Lemma 9 [9]. $M_{22}(\cdot) \leq N_r(\cdot) \leq M_r(\cdot)$.

7. Proofs of Theorems 1 and 2

First we shall prove Theorem 1. We can choose $p > 1$ such that $2p^2 - 1 < 1 + \varepsilon/2$. Set

$$(7.1) \quad \delta_n = \exp\{-p^n\}, \quad \varepsilon_n = \delta_{n-1}/6, \quad \varphi(x) = \sqrt{2(2p^2 - 1)N \log 1/x}, \\ I_n = \{(i, j, k) : k = n-1, n, \|t_i^{(n)} - t_j^{(n)}\| \leq \delta_n\} \quad \text{and} \\ E_{i,j}^n = \{\omega : |X(t_i^{(n)}) - X(t_j^{(n)})| \geq \sigma(\|t_i^{(n)} - t_j^{(n)}\|) \\ \times \varphi(\|t_i^{(n)} - t_j^{(n)}\|)\}, \quad ((i, j, k) \in I_n),$$

where $\{t_i^{(n)}, i=1, 2, \dots, N_{\varepsilon_n}(D)\}$ is a minimal ε_n -net of D . Then it follows from Lemma 4, (i) and Lemma 8 that

$$(7.2) \quad P(E_{i,j}^n) \leq c_{30} \frac{\delta_n^{(2p^2-1)N}}{\sqrt{\log 1/\delta_n}}$$

and

$$(7.3) \quad \#I_n \leq c_{31} \delta_n^N \varepsilon_n^{-2N} \leq c_{32} \delta_n^{1-2p^2N}.$$

Therefore combining (7.2) and (7.3) we have

$$(7.4) \quad \sum_n \sum_{(i,j,k) \in I_n} P(E_n^{i,j,k}) \leq \sum_n c_{30} \cdot c_{32} / \sqrt{\log 1/\delta_n} < +\infty.$$

Using Borel-Cantelli Lemma, there exists a integer $n_0(\omega)$, with probability one, such that

$$(7.5) \quad |X(t_i^{(n)}) - X(t_j^{(n)})| \leq \sigma(\|t_i^{(n)} - t_j^{(n)}\|) \varphi(\|t_i^{(n)} - t_j^{(n)}\|)$$

holds for any $(i, j, k) \in I_n$, $n \geq n_0$.

Denote $\{t_{i(n)}^{(n)}\}_{n \geq n_0}$ be a sequence such that $\|s - t_{i(n)}^{(n)}\| \leq 3\varepsilon_n$ for a fixed $s \in D$. From the inequality $\|t_{i(n+1)}^{(n+1)} - t_{i(n)}^{(n)}\| \leq \|t_{i(n+1)}^{(n+1)} - s\| + \|s - t_{i(n)}^{(n)}\| \leq 3\varepsilon_{n+1} + 3\varepsilon_n \leq 6\varepsilon_n = \delta_{n+2} < \delta_{n+1}$, we have $(i(n+1), i(n), n) \in I_{n-1}$. Hence it follows from (7.5) and monotonicity of $\sigma(x)\sqrt{\log 1/x}$ that

$$(7.6) \quad |X(s) - X(t_{i(n)}^{(n)})| \leq \sum_{k=n}^{\infty} |X(t_{i(k+1)}^{(k+1)}) - X(t_{i(k)}^{(k)})| \leq \sum_{k=n}^{\infty} \sigma(\delta_{k+2}) \varphi(\delta_{k+2}).$$

But there exists a $c_{33}(p)$ such that

$$(7.7) \quad \varphi(\delta_{k+2}) \leq c_{33}(p) (\sqrt{\log 1/\delta_{k+2}} - \sqrt{\log 1/\delta_{k+1}}),$$

therefore we have

$$(7.8) \quad \sum_{k=n}^{\infty} \sigma(\delta_{k+2}) \varphi(\delta_{k+2}) \leq c_{33}(p) G_{\sigma}(\sqrt{\log 1/\delta_{n+1}}).$$

While for any $s, t \in D$ satisfying $\delta_{n+1} \leq \|s - t\| \leq \delta_n$ ($n \geq n_0$), there exists $t_{i_1}^{(n)}, t_{i_2}^{(n)}, t_0$ such that

$$(7.9) \quad \|s - t_{i_1}^{(n)}\| \leq \varepsilon_n, \quad \|t_{i_1}^{(n)} - t_0\| = \|s - t\| - \varepsilon_n, \\ \|t_0 - t_{i_2}^{(n)}\| \leq \varepsilon_n.$$

Thus we have

$$(7.10) \quad |X(s) - X(t)| \leq |X(s) - X(t_{i_1}^{(n)})| + |X(t_{i_1}^{(n)}) - X(t_{i_2}^{(n)})| \\ + |X(t_{i_2}^{(n)}) - X(t)| \\ \leq \sigma(\|s - t\|) \varphi(\|s - t\|) + 2c_{33}(p) G_{\sigma}(\sqrt{\log 1/\|s - t\|}),$$

because of $\|t - t_{i_2}^{(n)}\| \leq 3\varepsilon_n$, $\|t_{i_1}^{(n)} - t_{i_2}^{(n)}\| \leq \|s - t\|$ and (7.8). This completes the proof of Theorem 1. This theorem is true without concavity of $\sigma^2(x)$ and is also true under the condition $\sqrt{E[(X(\bar{s}) - X(\bar{t}))^2]} \leq \sigma(\|s - t\|)$ instead of the equality.

Remark 1 is easily obtained by setting

$$\delta_n = \exp\{-H^{-1}(p^n)\} \quad \text{and} \quad \varphi(x) = \sqrt{(4+\varepsilon)H(\log 1/x)}$$

instead of (7.1). The last part of Remark 1 is obtained from the estimate $\#I_n \ll c_{34} \exp\{(1+\varepsilon)H(\log 1/\delta_n)\}$ ($n \rightarrow \infty$) by virtue of (A.1)**).

Next we shall prove Theorem 2. Denote by $\{t_i^{(n)}, i=1, 2, \dots, M_{34n}(D)\}$ a maximal $3\varepsilon_n$ -distinguishable set and choose $s_i^{(n)}$ such that $\|t_i^{(n)} - s_i^{(n)}\| = \varepsilon_n = 2^{-n}$. Set

$$(7.11) \quad F_{n,i} = \{\omega : |X(t_i^{(n)}) - X(s_i^{(n)})| \leq \sigma(\|t_i^{(n)} - s_i^{(n)}\|) \\ \times \varphi(\|t_i^{(n)} - s_i^{(n)}\|)\},$$

where $\varphi(x) = \sqrt{(1-\varepsilon)N \log 1/x}$.

Setting

$$\gamma_{i,j}^{(n)} = E[(X(t_i^{(n)}) - X(s_i^{(n)}))(X(t_j^{(n)}) - X(s_j^{(n)}))]/\sigma^2(\delta_n),$$

we have

$$(7.12) \quad \gamma_{i,j}^{(n)} \leq \sigma^2\left(\frac{4}{5}\varepsilon_n\right)/2\sigma^2(\varepsilon_n) \leq 1/2$$

from Lemma 6, (i), Lemma 7, (i) and concavity of $\sigma^2(x)$. Now define auxiliary random variables $Z_i = \xi + \eta_i$; $i=1, 2, \dots, M_{34n}(D)$, where $\{\xi, \eta_i, i=1, 2, \dots\}$ are mutually independent Gaussian random variables with $E[\xi^2] = E[\eta_i^2] = 1/2$ and $E[\xi] = E[\eta_i] = 0$. Then by Lemma 4, (v), Lemma 8 and choosing ε' such that $(1-\varepsilon)(1+\varepsilon)^2 < 1$, we have

$$(7.13) \quad \sum_i^{\infty} P(\cap_i F_{n,i}) \leq \sum_i^{\infty} P(Z_i \leq \varphi(\varepsilon_n)) \quad \text{for any } i \\ \leq \sum_i^{\infty} P(\xi \leq -\varepsilon'\varphi(\varepsilon_n)) + \sum_i^{\infty} P(\eta_i \leq (1+\varepsilon')\varphi(\varepsilon_n)) \quad \text{for any } i \\ < +\infty.$$

This implies that (4.7) holds with probability one, and if $\sigma(x)$ is a *n.r.v.f.*, choosing $k\varepsilon_n$ -distinguishable set such that $\gamma_{i,j}^{(n)} \ll \varepsilon'$, we can set $\varphi(x) = \sqrt{(2-\varepsilon)N \log 1/x}$ instead of (7.11). Remark 3 is obtained only by setting $\varphi(x) = \sqrt{(1-\varepsilon)H(\log 1/x)}$ in the proof of Theorem 2.

8. Proofs of Theorems 3 and 4

First we shall prove Theorem 3. For the proof of the first part the following lemma plays an essential role. The idea of this lemma is due to T. Sirao [17].

Lemma 9. *Let (S, ρ) be a compact metric space satisfying $N_\varepsilon(K) \leq c_{22}(d(K)/\varepsilon)^N$ for any compact subset K of S , and let $\{X(s); s \in S\}$ be a path continuous real valued G.p. with $E[X(s)] = 0$ and $E[X(s)^2] = 1$. Assume that there exists a non-decreasing continuous function $\sigma(x)$ such that $\sqrt{E[(X(s) - X(t))^2]} \leq \sigma(\rho(s, t))$ and $F_\sigma(1) < +\infty$. Set*

$$(8.1) \quad A = \{\omega : \sup_{s \in K} X(s) \geq x + d_\sigma F_\sigma(d(K))\}$$

for any compact subset K of S , where d_σ is a constant larger than $2^{N/2}(\sqrt{2} + 1)\sqrt{N}$, then there exists a constant d_σ independent of x and K such that

$$(8.2) \quad P(A) \leq d_\sigma \Phi(x)$$

holds for any

$$(8.3) \quad 0 < x \sigma(d(K)) \leq d_{\sigma_0},$$

where d_{σ_0} is an arbitrary constant independent of x and K .

Proof. Let $\{t_i^{(n)}; i = 1, 2, \dots, N_{\varepsilon_n}(K)\}$ be minimal ε_n -net of K and set

$$\varepsilon_n = d(K) \exp\{-2^{n+1}\}, \quad x_n = d_\sigma(\sqrt{2} - 1)2^{(n-1)/2} \sigma(\varepsilon_{n-1}),$$

$$A^* = \{\omega : \sup_{s \in K} X(s) > x + \sum_{k=1}^{\infty} x_k\},$$

$$A_n^{(i,j)} = \{\omega : X(t_i^{(j)}) > x + \sum_{k=1}^n x_k\},$$

$$A_n = \bigcup_{\substack{1 \leq i \leq N_{\varepsilon_i}(K) \\ 0 \leq j \leq n}} A_n^{(i,j)},$$

$$A_n(\infty) = \{\omega : \max_{\substack{1 \leq i \leq N_{\varepsilon_i}(K) \\ 0 \leq j \leq n}} X(t) > x + \sum_{k=1}^{\infty} x_k\}.$$

Since $\sum_{k=1}^{\infty} x_k < d_\sigma F_\sigma(d(K))$ and $A_n(\infty) \subset A_{n+1}(\infty) \subset \dots$,

$A^* = \bigcup_{n=1}^{\infty} A_n(\infty)$ and $A_n(\infty) \subset A_n$, we have

$$(8.4) \quad P(A) \leq P(A^*) \leq \liminf_{n \rightarrow \infty} P(A_n).$$

Now we can estimate $P(A_n)$ as follows.

$$(8.5) \quad \begin{aligned} P(A_n) &\leq P(A_{n-1}) + P(A_n \cap A_{n-1}^c) \\ &\leq P(A_{n-1}) + \sum_{i=1}^{N_{\varepsilon_n}(K)} P(A_n^{(n,i)} \cap A_{n-1}^c) \\ &\leq P(A_{n-1}) + \sum_{i=1}^{N_{\varepsilon_n}(K)} P(A_n^{(n,i)} \cap A_{n-1}^{(n-1, m(i))c}), \end{aligned}$$

where $m(i)$ is chosen such that $\rho(t_i^{(n)}, t_{m(i)}^{(n-1)}) \leq \varepsilon_{n-1}$.

Here we choose an auxiliary standard Gaussian random variable Y independent of $X(t_i^{(n)})$ such that $X(t_{m(i)}^{(n-1)}) = rX(t_i^{(n)}) + \sqrt{1-r^2}Y$, where $r = E[X(t_i^{(n)})X(t_{m(i)}^{(n-1)})]$. Since by definition

$$(8.6) \quad \begin{aligned} r &= 1 - \frac{1}{2} E[(X(t_i^{(n)}) - X(t_{m(i)}^{(n-1)}))^2] \\ &\geq 1 - \frac{1}{2} \sigma^2(\varepsilon_{n-1}) \equiv r_0, \end{aligned}$$

it follows from Lemma 4, (iv) that

$$(8.7) \quad \begin{aligned} P(A_n^{(n,i)} \cap A_{n-1}^{(n-1, m(i))c}) &\leq P(X(t_i^{(n)}) > x + \sum_{k=1}^n x_k, r_0 X(t_i^{(n)}) + \sqrt{1-r_0^2} \leq x + \sum_{k=1}^n x_k) \\ &\leq \Phi(x) \Phi((1-r_0)(x + \sum_{k=1}^n x_k) / \sqrt{1-r_0^2} - r_0 x_n / \sqrt{1-r_0^2}). \end{aligned}$$

Hence for any $n \geq n_0 = \min \{n : \sigma(\varepsilon_{n-1}) \leq 1\}$ we have

$$(8.8) \quad \begin{aligned} &\Phi((1-r_0)(x + \sum_{k=1}^n x_k) / \sqrt{1-r_0^2} - r_0 x_n / \sqrt{1-r_0^2}) \\ &\leq \Phi(x_n / 2\sigma(\varepsilon_{n-1}) - x\sigma(d(K))/2 - d_\sigma F_\sigma(d(K))/2) \\ &\leq \Phi((\sqrt{2}-1)d_\sigma 2^{(n-3)/2} - d_{10}/2 - d_\sigma F_\sigma(d(K))/2) \\ &= \Phi(P_n). \end{aligned}$$

where $P_n = (\sqrt{2}-1)d_\sigma 2^{(n-3)/2} - d_{10}/2 - d_\sigma F_\sigma(d(K))/2$.

Therefore combining (8.5), (8.7) and (8.8) we have

$$(8.9) \quad \begin{aligned} P(A_n) &\leq P(A_{n-1}) + N_{\varepsilon_n}(K) \Phi(P_n) \Phi(x) \\ &\quad \dots \dots \\ &\leq P(A_{n_0-1}) + \Phi(x) \sum_{k=n_0}^n N_{\varepsilon_k}(K) \Phi(P_k). \end{aligned}$$

Since $P(A_{n_0-1}) \leq \Phi(x) \sum_{k=0}^{n_0-1} N_{\varepsilon_k}(K)$ and there exists a constant d ,

independent of K such that

$$(8.10) \quad \sum_{k=0}^{n-1} N_{t,k}(K) + \sum_{k=n_0}^{\infty} N_{t,k}(K) \Phi(p_k) \leq d_5 < +\infty,$$

we obtain the final result (8.2).

Using this lemma we shall prove the first part of Theorem 3. It is sufficient to show by Lemma 2. (iii) that $\varphi \in \mathcal{U}^m(X)$ if $I_n(\sigma, \varphi) < +\infty$ under the assumption

$$(8.11) \quad \sqrt{3N \log 1/x} \gg \varphi \gg \sqrt{N \log 1/x}.$$

Let $S = (D \times D, \rho)$ be a direct product of D with metric $\rho((t, s), (t', s')) = \sqrt{\|t - t'\|^2 + \|s - s'\|^2}$. $(t, s), (t', s') \in D \times D$ and $\{t_i^{(m)}, i = 1, 2, \dots, N_n(D)\}$ be a minimal ε_n -net of D . Set

$$(8.12) \quad \begin{aligned} \varepsilon_n &= \sigma^{-1} \left(\frac{\sigma(2^{-n})}{\sqrt{n \log 2}} \right), & \delta_n &= 2^{-n-1} - 2\varepsilon_n \\ I_n &= \{(i, j) : \delta_{n-1} - 2\varepsilon_n \leq \|t_i^{(m)} - t_j^{(m)}\| \leq \delta_n + 2\varepsilon_n\} \\ D_i^{(m)} &= \{s \in D : \|t_i^{(m)} - s\| \leq \varepsilon_n\} \\ S_{i,j}^{(m)} &= D_i^{(m)} \times D_j^{(m)} \quad (i \neq j) \\ Y(s, t) &= \frac{X(s) - X(t)}{\sigma(\|s - t\|)}, \quad ((s, t) \in S_{i,j}^{(m)}) \\ E_{i,j}^{n,m} &= \{\omega : \sup_{(s,t) \in S_{i,j}^{(m)}} Y(s, t) > \varphi(2^{-n-1}) + 2d_5 F_\sigma(4\varepsilon_n) / \sigma(\delta_{n-1} - 4\varepsilon_n)\} \end{aligned}$$

Then it follows by concavity of $\sigma^2(x)$ that

$$(8.13) \quad \begin{aligned} E[(Y(s, t) - Y(s', t'))^2] \\ \leq \frac{2(\sigma^2(\|s - s'\|) - \sigma^2(\|t - t'\|))}{\sigma(\|s - t\|)\sigma(\|s' - t'\|)} \leq \frac{4\sigma^2(\rho((s, t), (s', t')))}{\sigma^2(\delta_{n-1} - 4\varepsilon_n)}. \end{aligned}$$

To apply Lemma 9 for $E_{i,j}^{n,m}$, we shall check the condition (8.3). Since $\lim_{n \rightarrow \infty} \varepsilon_n / \delta_{n-1} = 0$ and $d(S_{i,j}^{(m)}) \leq 2\sqrt{2}\varepsilon_n$, there exists an absolute constant d_{11} such that

$$(8.14) \quad 2\varphi(2^{-n-1}) \cdot \sigma(4\varepsilon_n) / \sigma(\delta_{n-1} - 4\varepsilon_n) \leq d_{11},$$

because of (8.11) and the property of *n.r.v.f.*. Hence using Lemma 9 we have

$$(8.15) \quad P(E_{i,j}^{n,m}) \leq d_{12} \Phi(\varphi(2^{-n-1})).$$

Since

$$(8.16) \quad \begin{aligned} \#I_n &\leq c_{36}(\delta_n + 2\varepsilon_n)^N \varepsilon_n^{-2N} \\ &\leq c_{37}(2^{-n} - 2^{-n-1})2^{-(n+1)(N-1)} \varepsilon_n^{-2N}, \end{aligned}$$

we have

$$(8.17) \quad \begin{aligned} &\sum_n \sum_{(i,j) \in I_n} P(E_{i,j}^n) \\ &\leq c_{38} \sum_n (2^{-n} - 2^{-n-1})2^{-(n+1)(N-1)} \exp \left\{ -\frac{1}{2} \varphi^2(2^{-n-1}) \right\} \\ &\quad \left[\sigma^{-1}(\sigma(2^{-n})/\sqrt{n \log 2}) \right]^{2N} \varphi(2^{-n-1}) \\ &\leq c_{39} I_u(\sigma, \varphi) < +\infty. \end{aligned}$$

As the same manner as (8.14) we have

$$(8.18) \quad 2d_s F_\sigma(4\varepsilon_n) \varphi(2^{-n-1})/\sigma(\delta_{n+1} - 4\varepsilon_n) \leq c_{39}.$$

This yields $\varphi + c_{39}/\varphi \in \mathcal{U}^*(X)$, but $I_u(\sigma, \varphi) < +\infty$ implies $I_u(\sigma, \varphi - c_{39}/\varphi) < +\infty$. Hence we have $\varphi \in \mathcal{U}^*(X)$ if $I_u(\sigma, \varphi) < +\infty$.

Next we shall prove the last part of Theorem 3. For this purpose we shall use Lemma 3, (ii) and notice that it is sufficient to prove under the assumption (8.11) by Lemma 2, (iv). Let us assume that $I_u(\sigma, \varphi) = +\infty$ and let $\{t_i^{(n)} : i=1, 2, \dots, M_{2\varepsilon_n}(D)\}$ be a maximal $2\varepsilon_n$ -distinguishable set of D . Set

$$\begin{aligned} \varepsilon_n &= \sigma^{-1} \left(\frac{\sigma(2^{-n-1})}{\sqrt{(n-1) \log 2}} \right), \\ L_n &= \{(i, j) : 2^{-n} \leq \|t_i^{(n)} - t_j^{(n)}\| \leq 2^{-n-1}\}, \end{aligned}$$

and

$$F_{i,j}^n = \{\omega : X(t_i^{(n)}) - X(t_j^{(n)}) > \sigma(\|t_i^{(n)} - t_j^{(n)}\|) \varphi(\|t_i^{(n)} - t_j^{(n)}\|)\}$$

for $(i, j) \in L_n$. Since we see

$$(8.19) \quad P(F_{i,j}^n) \geq \frac{\exp \left\{ -\frac{1}{2} \varphi^2(2^{-n}) \right\}}{2\sqrt{2\pi} \varphi(2^{-n})}$$

and

$$\#L_n \geq c_{41}(2^{-n} - 2^{-n-1})2^{-n(N-1)} \varepsilon_n^{-2N},$$

we have

$$(8.20) \quad \sum_n \sum_{(i,j) \in L_n} P(F_{i,j}^n) \geq$$

$$\begin{aligned} &\geq c_{12} \sum_{n=1}^{\infty} \frac{(2^{-n} - 2^{-n-1}) 2^{-n(N-1)} \exp \left\{ -\frac{1}{2} \varphi^2(2^{-n}) \right\}}{\varepsilon_n^{2N} \varphi(2^{-n})} \\ &\geq c_{12} I_u(\sigma, \varphi) = +\infty. \end{aligned}$$

To check the condition (b) of Lemma 3, the four point property plays an important role, which was pointed out by I. Kubo. Now we shall enumerate the events $\{F_{p,q}^n\}$ in some linear order $\{F^{(k)}\}_{k=1}^{\infty}$ such that $k > k'$ if $\|t_i^{(k)} - t_j^{(k)}\| \geq \|t_p^{(k')} - t_q^{(k')}\|$, where $F^{(k)}(F^{(k')})$ correspond to $F_{p,q}^n(F_{i,j}^n)$. Let us fix (i, j) and sufficiently large n , and let us write for simplicity that

$$\begin{aligned} \|t_i^{(n)} - t_j^{(n)}\| &= r_{i,j}, \quad \|t_p^{(m)} - t_q^{(m)}\| = r_{p,q}, \quad \|t_i^{(m)} - t_p^{(m)}\| = r_{i,p}, \\ \|t_i^{(n)} - t_q^{(m)}\| &= r_{i,q}, \quad \|t_j^{(n)} - t_p^{(m)}\| = r_{j,p}, \\ \text{and } \|t_j^{(n)} - t_q^{(m)}\| &= r_{j,q}. \end{aligned}$$

Now set

$$\begin{aligned} (8.21) \quad m_n &= n + \frac{3 \log n}{\alpha \log 2} \quad (\alpha \text{ is the exponent of n.r.v.f. } \sigma(x)), \\ \gamma_{p,q}^{(m)} &= E[(X(t_i^{(n)}) - X(t_j^{(n)}))(X(t_p^{(m)}) - X(t_q^{(m)}))] / \sigma(r_{i,j}) \sigma(r_{p,q}), \\ L_{m,1}^{(1)} &= \{(p, q) \in L_m : 1 - k/n \leq \gamma_{p,q}^{(m)} \leq 1 - (k-1)/n, 1 \leq k \leq \sqrt{n} + 1\}, \\ L_{m,1}^{(2)} &= \{(p, q) \in L_m : 1/\varphi(r_{i,j}) \varphi(r_{p,q}) \leq \gamma_{p,q}^{(m)} \leq 1 - 1/\sqrt{n}\}. \end{aligned}$$

Since we get by concavity of $\sigma^2(x)$

$$(8.22) \quad \gamma_{p,q}^{(m)} \leq \sigma(r_{p,q} \wedge r_{i,q} \wedge r_{j,p}) / \sigma(r_{i,j}),$$

combining (8.11), (8.22) and Lemma 5 we have

$$(8.23) \quad \gamma_{p,q}^{(m)} \varphi(r_{i,j}) \varphi(r_{p,q}) \leq \frac{\sigma(2^{-m_0-1}) \varphi(2^{-m_0-1}) \varphi(2^{-n+1})}{\sigma(2^{-n})} \ll 1, \\ \text{if } m \geq m_0.$$

Using Lemma 4. (ii) we have

$$(8.24) \quad P(F_{i,j}^n \cap F_{p,q}^m) \leq d_3 P(F_{i,j}^n) P(F_{p,q}^m)$$

for any $m \geq m_0$. On the other hand if $(p, q) \in L_{m,1}^{(1)}$, the following are satisfied whose proofs are given below :

$$(8.25) \quad (r_{i,p} + r_{j,p}) \vee (r_{i,q} + r_{j,q}) \leq d_{13} r_{i,j},$$

$$(8.26) \quad r_{p,q} \geq d_{14} r_{i,j},$$

$$(8.27) \quad r_{i,q} \geq r_{j,q}, \quad r_{j,p} \geq r_{i,p},$$

$$(8.28) \quad m \leq n + d_{15}, \text{ where } d_{15} \text{ is independent of } n.$$

By Lemma (6), (i) and (8.22) we have

$$(8.29) \quad \begin{aligned} \gamma_{p,q}^{(m)} &\leq \frac{\sigma^2(|r_{i,q} + r_{j,p} - r_{i,p} - r_{j,q}|)}{2\sigma(r_{i,j})\sigma(r_{p,q})} \\ &\leq \frac{\sigma^2(4r_{i,j}r_{p,q}/R)}{2\sigma(r_{i,j})\sigma(r_{p,q})}, \end{aligned}$$

where $R = (r_{i,p} + r_{j,p}) \vee (r_{i,q} + r_{j,q})$. This implies (8.25). It follows (8.26) from

$$(8.30) \quad \sqrt{3}/2 \leq \gamma_{p,q}^{(m)} \leq \frac{\sigma(r_{p,q})}{\sigma(r_{i,j})}.$$

Analogously we have

$$(8.31) \quad \sqrt{3}/2 \leq \gamma_{p,q}^{(m)} \leq \frac{\sigma^2(r_{i,q}) + \sigma^2(r_{i,j}) - \sigma^2(r_{j,q})}{\sqrt{3}\sigma^2(r_{i,j})}$$

which yields (8.27). It follows (8.28) from (8.26).

Now we shall estimate $\#L_m^{(1)}$, for $n \leq m \leq n + d_{15}$, $1 \leq k \leq \sqrt{n} + 1$. For this end we divide following four cases.

Case 1. $r_{i,q} \vee r_{j,p} \leq r_{i,j}$. Using concavity of $\sigma^2(x)$ we have

$$(8.32) \quad \begin{aligned} \gamma_{p,q}^{(m)} &\leq \frac{\sigma^2(r_{i,q}) + \sigma^2(r_{p,q}) - \sigma^2(r_{i,p})}{2\sigma(r_{p,q})\sigma(r_{i,q})} \\ &\leq 1 - \frac{\sigma^2(r_{i,p}) - \sigma^2(r_{p,q})(\sqrt{1 + \sigma^2(r_{i,p})/\sigma^2(r_{p,q})} - 1)^2}{2\sigma(r_{p,q})\sigma(r_{i,j})}. \end{aligned}$$

Since combining (8.26) and $r_{i,p} \leq r_{j,p} \leq r_{i,j}$ we have

$$(8.33) \quad \sqrt{1 + \sigma^2(r_{i,p})/\sigma^2(r_{p,q})} - 1 \leq c_{15}\sigma(r_{i,p})/\sigma(r_{p,q}) \\ (0 < c_{15} < 1),$$

there exists a constant $c_{16} > 0$ such that

$$(8.34) \quad \gamma_{p,q}^{(m)} \leq 1 - c_{16}\sigma^2(r_{i,p})/\sigma^2(r_{i,j}).$$

As the same manner as (8.34) we get

$$(8.35) \quad \gamma_{p,q}^{(m)} \leq 1 - c_{15}\sigma^2(r_{j,q})/\sigma^2(r_{i,j}).$$

Case 2. $r_{i,q} \wedge r_{j,p} \geq r_{i,j}$. First we shall assume that $r_{i,p} \leq r_{j,q}$.

By concavity of $\sigma^2(x)$ and Lemma 6, (ii) we have

$$\begin{aligned}
 (8.36) \quad \gamma_{p,q}^{(m)} &\leq \frac{\sigma^2(r_{i,q}) + \sigma^2(r_{j,p}) - \sigma^2(r_{i,p}) - \sigma^2(r_{j,q})}{2\sigma^2(r_{p,q})} \\
 &\leq 1 - \frac{\sigma^2(r_{i,p}) + \sigma^2(r_{j,q}) - \sigma^2(r_{i,q}) - \sigma^2(r_{j,p}) + 2\sigma^2(r_{p,q})}{2\sigma^2(r_{p,q})} \\
 &\leq 1 - \frac{\sigma^2(r_{i,p} + r_{p,q} - r_{j,p}) + \sigma^2(r_{j,q} + r_{p,q} - r_{i,q})}{2\sigma^2(r_{i,j})}.
 \end{aligned}$$

We shall show that it is impossible to find $d_{16} < \frac{d_{11}}{1+d_{11}}$ such that

$$(8.37) \quad r_{i,p} + r_{p,q} - r_{j,p} \leq d_{16} r_{i,p}$$

and

$$(8.38) \quad r_{j,q} + r_{p,q} - r_{i,q} \leq d_{16} r_{i,p}.$$

Because, combining (8.37), (8.48) and from Lemma 7, (ii) we have

$$\begin{aligned}
 (8.39) \quad 2(r_{p,q} - d_{16} r_{i,p}) &\leq r_{i,q} + r_{j,p} - r_{j,q} - r_{i,p} \\
 &\leq \frac{2r_{i,j} r_{p,q}}{r_{i,j} + r_{i,p}}.
 \end{aligned}$$

This implies by (8.25) that

$$\begin{aligned}
 (8.40) \quad r_{p,q} &\leq d_{16}(r_{i,j} + r_{i,p}) \\
 &\leq d_{16}(1 + d_{13})r_{i,j} < d_{14} r_{i,j},
 \end{aligned}$$

which contradicts to (8.26). Therefore we have

$$(8.41) \quad \gamma_{p,q}^{(m)} \leq 1 - \frac{\sigma^2(d_{16} r_{i,p})}{2\sigma^2(r_{i,j})}.$$

To estimate $r_{j,q}$, it needs to get the sharper inequality than (8.36). That is, we have

$$\begin{aligned}
 (8.42) \quad \gamma_{p,q}^{(m)} &\leq 1 - \frac{1}{2\sigma(r_{i,j})\sigma(r_{p,q})} \left\{ \sigma^2(r_{i,p}) + \sigma^2(r_{p,q}) - \sigma^2(r_{i,q}) \right. \\
 &\quad \left. + \sigma^2(r_{j,q}) + \sigma^2(r_{i,j}) - \sigma^2(r_{j,p}) - (\sigma(r_{i,j}) - \sigma(r_{p,q}))^2 \right\}.
 \end{aligned}$$

Using concavity of $\sigma^2(x)$ and Lemma 6, (ii) we have

$$(8.43) \quad \sigma^2(r_{i,p}) + \sigma^2(r_{p,q}) - \sigma^2(r_{i,q}) \geq \sigma^2(r_{p,q}) - \sigma^2(r_{i,q} - r_{i,p}) \geq 0,$$

$$(8.44) \quad \sigma^2(r_{j,q}) + \sigma^2(r_{i,j}) - \sigma^2(r_{j,p}) \geq \sigma^2(r_{j,p} + r_{i,j} - r_{j,p}),$$

and

$$(8.45) \quad (\sigma(r_{i,j}) - \sigma(r_{p,q}))^2 \leq \sigma^2(r_{p,q}) \left(\sqrt{1 + \frac{\sigma^2(r_{i,j} - r_{p,q})}{\sigma^2(r_{p,q})}} - 1 \right)^2 \\ \leq c_{46} \sigma^2(r_{i,j} - r_{p,q}) \quad (c_{46} < 1).$$

Hence combining (8.42), (8.43), (8.44) and (8.45) we have

$$(8.46) \quad \gamma_{n,q}^{(m)} \leq 1 - \frac{\sigma^2(r_{j,q} + r_{i,j} - r_{j,p}) - c_{46} \sigma^2(r_{i,j} - r_{p,q})}{2\sigma^2(r_{i,j})} \\ \leq 1 - \frac{(1 - c_{46}) \sigma^2(r_{j,q} + r_{i,j} - r_{j,p})}{2\sigma^2(r_{i,j})} \\ \leq 1 - \frac{(1 - c_{46}) \sigma^2(r_{j,q} - r_{i,p})}{2\sigma^2(r_{i,j})}, \quad \text{if } r_{j,q} \geq 2r_{i,p}.$$

If $r_{i,p} > r_{j,q}$, we have the same estimate as (8.41) and (8.46) by substitution of (i, p) and (j, q) .

Case 3. $r_{i,q} \leq r_{i,j} \leq r_{j,p}$. We notice that (8.34) and (8.46) are still valid without any change.

Case 4. $r_{j,p} \leq r_{i,j} \leq r_{i,q}$. This case is reduced to Case 3 by substitution of (i, q) and (j, p) .

Therefore through the all case it follows from Lemma 8 and Lemma 5 that

$$(8.47) \quad \#L_{m,b}^{(1)} \leq c_{47} (\sigma^{-1}(c_{46} \sqrt{k/n} \sigma(2^{-n})))^{2N} \varepsilon_m^{-2N} \leq c_{49} k^{\varepsilon_{50}}.$$

Next we shall estimate $\#L_m^{(2)}$ ($m \leq m_0$). It follows that

$$(8.48) \quad R \leq r_{i,j} (\log 1/r_{i,j})^{2/a},$$

where R is defined in (8.29). Because, if $(p, q) \in L_m^{(2)}$, $m \leq m_0$ then by (8.29) we have

$$\gamma_{r,q}^{(m)} \vee \overline{nm} \leq c_{51} \frac{\sigma^2(4r_{p,q} (\log 1/r_{i,j})^{-2/a}) \vee \overline{nm}}{\sigma^2(r_{p,q})} \ll 1.$$

Therefore we have

$$(8.49) \quad \#L_m^{(2)} \leq c_{52} (r_{i,j} (\log 1/r_{i,j})^{2/a})^{2N} \varepsilon_m^{-2N} \leq c_{53} n^{\varepsilon_{54}}.$$

Hence it follows by (8.11), (8.47), (8.49) and Lemma 4, (iii) that

$$\begin{aligned} & \sum_{m=0}^{n+d_1} \sum_{k=1}^{n+1} \sum_{(\phi, q) \in L_{\alpha}^{(k)}} P(F_{i,j}^n \cap F_{n,q}^m) + \sum_{m=0}^{m_0} \sum_{(\phi, q) \in L_{\alpha}^{(2)}} P(F_{i,j}^n \cap F_{n,q}^m) \\ & \leq [d_1 \sum_{k=1}^n c_k k^{-\infty} \exp\{-c_{55}k\} + c_{55}n^{c_{54}} \exp\{-c_{56}\sqrt{n}\}] P(F_{i,j}^n) \\ & \leq c_{57} P(F_{i,j}^n). \end{aligned}$$

Using Lemma 3, we have

$$(8.50) \quad P(\overline{\lim} F^{(k)}) \geq 1/d_3.$$

On the other hand we notice that by setting

$$(8.51) \quad \psi(x) = G^{-1} \left[G(\varphi(x)) + \log \int_x \frac{u^{N-1} \exp\left\{-\frac{1}{2} \varphi^2(u)\right\}}{\left[\sigma^{-1}\left(\frac{\sigma(u)}{\sqrt{\log 1/u}}\right)\right]^{2N} \varphi(u)} du \right],$$

where $G(x) = \frac{1}{2}x^2 + \log x$, we get $I_u(\sigma, \psi) = +\infty$ and $\lim_{x \rightarrow 0^+} \psi^2(x) - \varphi^2(x) = +\infty$. Therefore by (8.50), Lemma 1 and concavity of $\sigma^2(x)$, it follows that $\sqrt{\varphi^2(x) + c_{58}(\psi^2(x) - \varphi^2(x))} \in \mathcal{L}^u(X)$ for some $c_{58} > 0$, this yields $\varphi(x) \in \mathcal{L}^u(X)$. Thus we get the proof of the last half of Theorem 3.

The first part of Corollary 3 is an immediate consequence of Theorem 3. To prove the last part of Corollary 3, we set $f(x) = \int_0^x (\alpha - a(e^{-u})) du$. Since we can write

$$(8.52) \quad \sigma_2(x) = c(x) \exp\{-f(\log 1/x)\},$$

where $c(x)$ is a continuous functions such that $c_{59} \geq c(x) \geq c_{60} > 0$, we have

$$\begin{aligned} (8.53) \quad \sigma_2^{-1}\left(\frac{\sigma_2(x)}{\sqrt{\log 1/x}}\right) &= \exp\left\{-f^{-1}\left(f(\log 1/x) + \frac{1}{2} \log_{c_2} 1/x + c^*(x)\right)\right\} \\ &\asymp \exp\left\{-\left(\log 1/x + \frac{\log_{c_2} 1/x}{2\alpha}\right)\right\} \exp\left\{\frac{a(x) \log_{c_2} 1/x}{\alpha(\alpha - a(x))}\right\} \\ &\asymp \sigma_1^{-1}\left(\frac{\sigma_1(x)}{\sqrt{\log 1/x}}\right), \end{aligned}$$

if $|a(x) \log_{c_2} 1/x| \leq c_{17}$, which proves the last part of Corollary 3. Corollary 4 follows from the fact that there exists x_0 such that

$$\begin{aligned} (8.54) \quad & f^{-1}\left(f(\log 1/x) + \frac{1}{2} \log_{c_2} 1/x + c^*(x)\right) \\ &= \log 1/x + \frac{\log_{c_2} 1/x + 2c^*(x)}{2(\alpha - a(x_0))}. \end{aligned}$$

where, $\log 1/x \leq x_0 \leq f^{-1}(f(\log 1/x) + \frac{1}{2} \log_{(2)} 1/x + c(x))$, and that we can chose c_n by virtue of (4.14) such that

$$I_n(\sigma_1, \varphi) < +\infty, \quad I_n(\sigma_2, \varphi) = +\infty \quad \text{or} \quad I_n(\sigma_1, \varphi) = +\infty, \\ I_n(\sigma_2, \varphi) < +\infty \quad \text{for}$$

$$(8.55) \quad \varphi(x) = \sqrt{2N \log 1/x + \left(\frac{2N}{\alpha} + 1\right) \log_{(2)} 1/x + c_n \log_{(2)} 1/x}.$$

Finally we shall prove Theorem 4. The proof of this theorem is essentially based on the same method as that of Theorem 3. So we shall use the same notation and denote by (\cdot) , or (\cdot') if we need slightly modified notations. Set

$$h(x) = \int_0^x a(e^{-u}) du,$$

$$\bar{\varepsilon}_n = \exp \left\{ -h^{-1}(h(n \log 2) + \frac{1}{2} \log(n \log 2)) \right\},$$

$$\varphi_\varepsilon(x) = \sqrt{2N \log 1/x + 2N(1 + \varepsilon)(\log_{(2)} 1/x)/a(x)}.$$

Since by the condition (i) it follows that

$$(8.56) \quad h^{-1}(h(\log 1/x) + \frac{1}{2} \log_{(2)} 1/x + c_n) \\ \ll \log 1/x + \frac{(1 + \varepsilon') \log_{(2)} 1/x}{2a(x)}$$

for any $\varepsilon > \varepsilon' > 0$, we have $\varphi_\varepsilon(2^{-n-1})\sigma(4\bar{\varepsilon}_n)/\sigma(\delta_{n+1} - 4\bar{\varepsilon}_n) \leq d_{17}$. Hence using Lemma 9 we get

$$(8.57) \quad P(E_{i,j}^n) \leq d_{18} \Phi(\varphi_\varepsilon(2^{-n-1})).$$

Since

$$(8.58) \quad \#I_n \leq c_n(2^{-n} - 2^{-n-1})2^{-(n-1)N} \bar{\varepsilon}_n^{2N}$$

it follows that

$$(8.59) \quad \sum_{i,j}^{\infty} \sum_{i \in I_n} P(E_{i,j}^n) \leq c_n \sum_n (2^{-n} - 2^{-n-1}) 2^{-(n-1)N} \\ \times \exp \left\{ 2N \left(\log 2^n + \frac{1 + \varepsilon'}{2} \frac{\log_{(2)} 2^n}{a(2^{-n})} - \frac{1}{2} \varphi_\varepsilon^2(2^{-n-1}) \right) \right\} / \varphi_\varepsilon(2^{-n-1}) \\ < +\infty.$$

While from Lemma 11 we have

$$(8.60) \quad \frac{F_{\sigma}(4\bar{\varepsilon}_n) \varphi_{\varepsilon}(2^{-n-1})}{\sigma(\delta_{n+1} - 4\bar{\varepsilon}_n)} \leq \frac{c_{63}}{\sqrt{a(2^{-n-1})}}.$$

This yields $\varphi_{\varepsilon}(x) + c_{63}/(\varphi_{\varepsilon}(x)\sqrt{a(x)}) \ll \varphi_{2\varepsilon} \in \mathcal{U}^n(X)$, for any $\varepsilon > 0$, then the first part of Theorem 4 was proved.

Next we shall prove the last part of Theorem 4. Set

$$\varepsilon'_n = \exp \left\{ -h^{-1}(h(n \log 2) + \frac{1+\varepsilon}{2} \log(n \log 2)) \right\} \quad (\varepsilon < 0).$$

Since by the condition (i) it follows that

$$(8.61) \quad \begin{aligned} & \exp \left\{ -\left(n \log 2 + \frac{(1+\varepsilon) \log(n \log 2)}{2a(2^{-n})} \right) \right\} \\ & \gg \varepsilon'_n \gg \exp \left\{ -\left(n \log 2 + \frac{(1-\varepsilon) \log(n \log 2)}{2a(2^{-n})} \right) \right\} \quad (\varepsilon < 0), \end{aligned}$$

we get

$$(8.62) \quad \begin{aligned} & \sum_{i=1}^n \sum_{(i,j) \in E_n} P(F_{i,j}^n) \\ & \geq c_{67} \sum_{i=1}^n (2^{-n} - 2^{-n-1}) 2^{-n-N} \exp \left\{ 2N(\log 2^n + \frac{1+\varepsilon}{2} \frac{\log(2^n)}{a(2^{-n})}) \right\} \\ & \times \exp \left\{ -\frac{1}{2} \varphi_i^2(2^{-n-1}) \right\} / \varphi_i(2^{-n}) = +\infty \quad (\varepsilon < 0). \end{aligned}$$

To check the condition (b) of Lemma 3, we shall follow step by step the last part of the proof of Theorem 3. Set

$$(8.63) \quad \begin{aligned} h(x, y) &= \int_x^y a(e^{-u}) du, \\ \bar{m}_0 &= n + \frac{2 \log n}{a(2^{-n}) \log 2}. \end{aligned}$$

Since by the condition (ii) we get

$$(8.64) \quad \frac{\log_{(x)} 1/x}{a(t)} \ll -\varepsilon \log 1/x \quad (\varepsilon < 0),$$

it follows analogously in (8.23) that

$$(8.65) \quad \begin{aligned} & \gamma_{n-\sigma}^{(m)} \varphi_{\varepsilon}(r_{i,j}) \varphi_{\varepsilon}(r_{p,q}) \\ & \ll c_{69} \sigma(2^{-n-1}) \sqrt{\bar{m}n} / \sigma(2^{-n}) \end{aligned}$$

$$\begin{aligned} &\ll c_{60} n \exp \left\{ -h \left(n \log 2, n \log 2 + \frac{2 \log n}{a(2^{-n})} \right) \right\} \\ &\ll 1, \quad \text{if } m \geq \tilde{m}_0. \end{aligned}$$

By virtue of (8.61), we have

$$\begin{aligned} (8.66) \quad \#L_{m, k}^{(1)} &\leq c_{70} (\sigma^{-1}(c_{71} \sqrt{k/n} \sigma(2^{-n})/\varepsilon'_m))^{2N} \\ &\leq c_{72} \exp \{N(1+\varepsilon) \log k/a(2^{-n}) + N\varepsilon \log(n \log 2)/2a(2^{-n})\} \end{aligned}$$

for $1 \leq k \leq n^{1-\beta/2} + 1$, (β is defined in Condition (ii)). Therefore we get

$$\begin{aligned} (8.67) \quad &\sum_{m=n}^{n-d_{15}} \sum_{k=1}^{n^{1-\beta/2}+1} \sum_{(p, q) \in L_{m, k}^{(1)}} P(F_{i, j}^n \cap F_{p, q}^n) \\ &\leq c_{73} \sum_{k=1}^{n^{1-\beta/2}+1} \exp \{N(1+\varepsilon) \log k/a(2^{-n}) + N\varepsilon \log(n \log 2)/2a(2^{-n}) - c_{74} k\} \\ &\leq d_{19} P(F_{i, j}^n) \quad (\varepsilon < 0). \quad \times P(F_{i, j}^n) \end{aligned}$$

While if $R = (r_{i, p} + r_{j, p}) \vee (r_{i, p} + r_{j, q}) \geq 4 \exp \left\{ -n \log 2 + \frac{c_{75} \log n}{a(2^{-n})} \right\}$ and $m \leq m_0$, it follows by (8.29) that

$$\begin{aligned} (8.68) \quad &\gamma_{p, q}^{(m)} \varphi_\varepsilon(r_{i, j}) \varphi_\varepsilon(r_{p, q}) \\ &\leq \sigma(2^{-n-c_{75} \log n/a(2^{-n})}) \varphi_\varepsilon(2^{-n-1}) \varphi_\varepsilon(2^{-m_0-1})/2\sigma(2^{-n}) \\ &\leq c_{16} n \exp \{ -h((n+c_{75} \log n/a(2^{-n})) \log 2) - h(n \log 2) \} \\ &\ll 1 \end{aligned}$$

for a suitably chosen constant c_{75} , and we have

$$\begin{aligned} (8.69) \quad \#L^{(2)} &\leq c_{77} (\exp \{ -n \log 2 + c_{75} \log n/a(2^{-n}) \} / \varepsilon'_{m_0})^{2N} \\ &\leq c_{78} \exp \{ c_{79} \log n/a(2^{-n}) \} \\ &\leq c_{78} \exp \{ c_{80} n^{1-\beta} \log n \}. \end{aligned}$$

Accordingly we get by Lemma 4, (iii),

$$\begin{aligned} (8.70) \quad &\sum_{m=n}^{\tilde{m}_0} \sum_{(p, q) \in L_m^{(2)}} P(F_{i, j}^n \cap F_{p, q}^n) \\ &\leq \{ c_{78} 2 \log n \exp \{ c_{80} n^{1-\beta} \log n - c_{81} n^{1-\beta/2} \} / a(2^{-n}) \log 2 \} P(F_{i, j}^n) \\ &\leq d_{21} P(F_{i, j}^n). \end{aligned}$$

Hence combining (8.62), (8.67) and (8.70) we have $\varphi_{21} \in \mathcal{L}^n(X)$ for $\varepsilon < 0$. This completes the proof of Theorem 4.

9. Proofs of Theorems 5 and 6

The proofs of Theorems 5 and 6 are based on those of M. B. Marcus' theorems [13]. To prove Theorem 5 we shall use the following Lemma due to X. Fernique. Let (S, ρ) be a compact metric space such that

$$(9.1) \quad N_\varepsilon(K) \leq d_{22} \left(\frac{d(K)}{\varepsilon} \right)^N$$

for any compact subset K of S , and $\{X(s) : s \in S\}$ be a path continuous G.p. with $E[X(s)] = 0$ such that $E[(X(s) - X(t))^2] \leq \sigma^2(\rho(s, t))$, where $\sigma(x)$ is a non-decreasing continuous function satisfying $F_\sigma(1) < +\infty$.

Lemma 10. [5]

$$P(\sup_{s \in K} |X(s)| \geq x(\|\Gamma\|_K + 4F_\sigma(d(K))/\sqrt{\log \bar{p}})) \leq 9d_{22} p^{2N} \Phi(x)$$

for any $p > 1$ and $x \geq \sqrt{1 + 4N \log \bar{p}}$; where $\|\Gamma\|_K = \sup_{s, t \in K} |E[X(s)X(t)]|$.

Now we shall prove Theorem 5. Set

$$\begin{aligned} \varphi(x) &= \sqrt{2} \{ (\log_{22} 1/x) \sqrt{F_\sigma(x)/\sigma(x)} \}, \\ \delta_n &= e^{-n}, \quad S_n = \{s \in S : \delta_{n+1} \leq \|s - s_0\| \leq \delta_n\} \\ Y(s) &= \frac{X(s) - X(s_0)}{\sigma(\|s - s_0\|)}, \quad s \in S_n. \end{aligned}$$

Since we get

$$(9.2) \quad \begin{aligned} E[Y(s)^2] &= 1, \\ E[(Y(s) - Y(t))^2] &\leq \frac{\sigma^2(\|s - t\|)}{\sigma(\|s - s_0\|)\sigma(\|t - s_0\|)} \\ &\leq \sigma^2(\|s - t\|)/\sigma^2(\delta_{n+1}) \quad \text{for } s, t \in S_n, \end{aligned}$$

using Lemma 10 for $p^n = n$ it follows that

$$(9.3) \quad \begin{aligned} P(\sup_{s \in S_n} |Y(s)| \geq c_{22} \varphi(\delta_n)) \\ \leq 9d_{22} n^{2N/n} \Phi(x_n), \end{aligned}$$

where

$$x_n = \frac{c_{62} \sqrt{\log n} \varphi(\delta_n)}{\sqrt{\log n} + 4\sqrt{q} F_\sigma(2\delta_n)/\sigma(\delta_{n+1})}.$$

By the definition of φ we have

$$(9.4) \quad x_n \geq \frac{c_{62} \sqrt{2 \log n}}{1 + 4\sqrt{q} c_{63}}.$$

Hence for suitably chosen q , c_{62} and c_{63} we have

$$(9.5) \quad \sum_n P(\sup_{t \in S_n} |Y(t)| \geq c_{62} \varphi(\delta_n)) < +\infty.$$

In particular in case of (i) we can set c_{62} arbitrarily closed to 1 because c_{63} can be chosen arbitrarily small for large n . The proof follows from Lemma 12 and Lemma 3.

Next we shall prove (i) of Theorem 6. Set

$$\begin{aligned} \varphi_t(x) &= \sqrt{(2-\varepsilon) \log_{(2)} \bar{1}/x}, \\ \|e\| &= 1, \quad s^n = s_0 + 2^{-n} e \in D \\ \gamma_{n,m} &= E[(X(s_n) - X(s_t))(X(s_m) - X(s_0))]/\sigma(2^{-n})\sigma(2^{-m}) \\ F^n &= \{\omega : |X(s_n) - X(s_0)| \geq \sigma(2^{-n})\varphi_t(2^{-n})\}. \end{aligned}$$

Then we have

$$(9.6) \quad \begin{aligned} \sum_n P(F^n) &= \sum_n \Phi(\varphi_t(2^{-n})) \\ &= \sum_n \Phi(\sqrt{(2-\varepsilon) \log(n \log 2)}) = +\infty. \end{aligned}$$

Since it follows that

$$(9.7) \quad \gamma_{n,m} = \frac{\sigma^2(2^{-n}) + \sigma^2(2^{-m}) - \sigma^2(2^{-n} - 2^{-m})}{2\sigma(2^{-n})\sigma(2^{-m})} \leq \sigma(2^{-m})/\sigma(2^{-n}),$$

we have

$$\begin{aligned} &\gamma_{n,m} \varphi_t(2^{-n}) \varphi_t(2^{-m}) \\ &\leq \sigma(2^{-m})(2-\varepsilon) \sqrt{\log(n \log 2) \log(m \log 2)}/\sigma(2^{-n}) \\ &\ll 1 \quad \text{for } m \geq m_0. \end{aligned}$$

This follows from Lemma 4. (ii) that

$$(9.8) \quad P(F^n \cap F^m) \leq d_3 P(F^n) P(F^m).$$

Since there exists a constant c_{64} such that $\gamma_{n,m} \leq \sigma(2^{-m})/\sigma(2^{-n})$

$\leq 1/2$ for $m \geq n + c_{34}$, we have

$$(9.9) \quad \left(\sum_{n=n_0}^n + \sum_{n=n_0}^{m_0} \right) P(F^n \cap F^m) \\ \leq c_{34} P(F^n \cap F^m) + \frac{3 \log n}{\alpha \log 2} \exp \{-c_{35} \log n\} P(F^n) \\ \leq d_{23} P(F^n).$$

Combining (9.6), (9.8) and (9.9) we get Theorem 6, (i). To prove Theorem 6, (ii) set

$$\varphi(x) = \sqrt{1/a(x)}, \quad r_n = \inf \{r; a(r) = 1/\log n\} \\ F^n = \{\omega; X(s_0 + r_n e) - X(s_0 + r_{n+1} e) \geq c_{36} \sigma(r_n) \varphi(r_n)\}.$$

Then we have

$$(9.10) \quad P(F^n) = \Phi \left(\frac{c_{36} \sigma(r_n) \varphi(r_n)}{\sigma(r_n - r_{n+1})} \right).$$

By the definition of r_n and (5.7), we get

$$(9.11) \quad \frac{1}{\log n} - \frac{1}{\log(n+1)} = a(r_n) - a(r_{n+1}) \\ \leq c_{24} \frac{r_n - r_{n+1}}{r_{n+1}} a(r_{n+1}) \\ \leq \frac{c_{24}}{\log(n+1)} \frac{r_n - r_{n+1}}{r_{n+1}}.$$

This follows

$$(9.12) \quad \frac{r_{n+1}}{r_n - r_{n+1}} \leq 2c_{24} n \log n.$$

While if $r_{n+1} \leq r_n/2$ we have

$$(9.13) \quad \frac{\sigma^2(r_n)}{\sigma^2(r_n - r_{n+1})} \leq \frac{\sigma^2(r_n)}{\sigma^2(r_n/2)} \leq c_{37},$$

and if $r_{n+1} \geq r_n/2$ we get by virtue of (9.12)

$$(9.14) \quad \frac{\sigma^2(r_n)}{\sigma^2(r_n - r_{n+1})} \leq c_{38} \exp \left\{ 2 \int_{r_n - r_{n+1}}^{r_n} a(s)/s ds \right\} \\ \leq c_{38} \exp \left\{ \frac{2}{\log n} \log \frac{2r_{n+1}}{r_n - r_{n+1}} \right\} \\ \leq c_{39}.$$

Hence combining (9.10), (9.13) and (9.14) it follows by choosing c_{66} suitably that

$$(9.15) \quad \sum_n P(F^n) = \sum_n \Phi(c_{66} \sqrt{\log n}) = +\infty,$$

while by concavity of $\sigma^2(x)$ we have

$$(9.16) \quad P(F^n \cap F^m) \leq P(F^n)P(F^m).$$

These imply that

$$(9.17) \quad \overline{\lim}_{n \rightarrow \infty} \frac{X(s_0 + r_n e) - X(s_0 + r_{n+1} e)}{\sigma(r_n) \varphi(r_n)} \geq c_{66}$$

with probability one. By the definition of $\varphi(x)$ and r_n we see

$$(9.18) \quad \lim_{n \rightarrow \infty} \frac{\varphi(r_n)}{\varphi(r_{n-1})} = \lim_{n \rightarrow \infty} \sqrt{\frac{\log(n+1)}{\log n}} = 1.$$

Therefore it follows from (9.17), (9.18) and symmetricity of $X(s)$ that

$$(9.19) \quad \overline{\lim}_{n \rightarrow \infty} \frac{X(s_0 + r_n e) - X(s_0)}{\sigma(r_n) \varphi(r_n)} \geq c_{66}/2$$

or

$$(9.20) \quad \overline{\lim}_{n \rightarrow \infty} \frac{X(s_0 + r_{n+1} e) - X(s_0)}{\sigma(r_n) \varphi(r_n)} \geq c_{66}/2$$

with probability one. Thus we have

$$(9.21) \quad \overline{\lim}_{\|s - s_0\| \rightarrow 0} \frac{X(s) - X(s_0)}{\sigma(\|s - s_0\|) \varphi(\|s - s_0\|)} \geq c_{66}/2.$$

Analogously, by choosing $r_n = 2^{-n}$, we have

$$(9.22) \quad \overline{\lim}_{\|s - s_0\| \rightarrow 0} \frac{X(s) - X(s_0)}{\sigma(\|s - s_0\|) \sqrt{\log_{(2)} 1/\|s - s_0\|}} \geq c_{66}$$

with probability one. This completes the proof of Theorem 6, (ii).

10. Proofs of Theorems 7 and 8

Now we shall prove Theorems 7 and 8 by means of following the proofs of Theorems 3 and 4 step by step. An alogous

consideration in case of Theorem 3, implies that it is sufficient to prove Theorem 7 for $\varphi(x)$ such that

$$(10.1) \quad \sqrt{\log_{e_2} 1/x} \ll \varphi(x) \ll \sqrt{3 \log_{e_2} 1/x}.$$

First we assume that $I_1(\sigma, \varphi) < +\infty$. One sets

$$(10.2) \quad \begin{aligned} \varepsilon_n &= \sigma^{-1} \left(\frac{\sigma(2^{-n})}{\sqrt{\log(n \log 2)}} \right) \\ T_n &= \{s \in D; 2^{-n-2} \leq \|s - s_0\| \leq 2^{-n-1}\}. \end{aligned}$$

Let $\{t_i^{(n)}, i=1, \dots, N_n(T_n)\}$ be a minimal ε_n -net of T_n , and set

$$\begin{aligned} S_{n,i} &= \{s \in T_n; \|t_i^{(n)} - s\| \leq \varepsilon_n\}, \\ E_n^? &= \{\omega : \sup_{s \in S_{n,i}^?} Y(s) \geq \varphi(2^{-n-1}) + d_6 F_\sigma(2\varepsilon_n) / \sigma(2^{-n-2} - \varepsilon_n)\}, \\ Y(s) &= \frac{X(s) - X(s_0)}{\sigma(\|s - s_0\|)}, \quad s \in S_{n,i}. \end{aligned}$$

Then we have

$$(10.3) \quad \begin{aligned} E[(Y(s) - Y(t))^2] &\leq \frac{\sigma^2(\|s - t\|)}{\sigma(\|s - s_0\|) \sigma(\|t - s_0\|)} \\ &\leq \sigma^2(\|s - t\|) / \sigma^2(2^{-n-2} - \varepsilon_n) \end{aligned}$$

and

$$(10.4) \quad \frac{\varphi(2^{-n-1}) \sigma(2\varepsilon_n)}{\sigma(2^{-n-2} - \varepsilon_n)} \leq d_{21}.$$

Therefore using Lemma 9 we get

$$(10.5) \quad P(E_n^?) \leq d_{23} \Phi(\varphi(2^{-n-1})).$$

Since

$$(10.6) \quad N_n(T_n) \leq c_{91} (2^{-n-1} / \varepsilon_n)^N,$$

we have

$$(10.7) \quad \begin{aligned} \sum_{i=1}^N \sum_{i=1}^{N_n(T_n)} P(E_n^?) &\leq \sum_{i=1}^N c_{92} (2^{-n} - 2^{-n-1}) \varepsilon_n^{-N} 2^{-(n+1)(N-1)} \\ &\quad \times \Phi(\varphi(2^{-n-1})) \\ &\leq c_{93} I_1(\sigma, \varphi) < +\infty. \end{aligned}$$

This yields $\varphi \in \mathcal{U}^l(X)$ by the same reason as the proof of Theorem 3.

Next we assume that $I_f(\sigma, \varphi) = +\infty$, and one sets

$$(10.8) \quad \varepsilon_n = \sigma^{-1} \left(\frac{\sigma(2^{-n})}{\sqrt{\log(n \log 2)}} \right) \\ S_n = \{s \in D; 2^{-n-1} \leq \|s - s_0\| \leq 2^{-n+2}\}.$$

Let $\{s_i^{(n)}; i=1, 2, \dots, M_n(S_n)\}$ be a maximal ε_n -distinguishable set of S_n and set

$$\|s_i^{(n)} - s_0\| = r_i^{(n)}, \\ F_n^* = \{\omega; X(s_i^{(n)}) - X(s_0) \geq \sigma(r_i^{(n)})\varphi(r_i^{(n)})\}.$$

Then we have

$$(10.9) \quad P(F_n^*) \geq \Phi(\varphi(2^{-n+1})),$$

and

$$(10.10) \quad M_n(S_n) \leq c_{95}(2^{-n}/\varepsilon_n)^N.$$

Combining (10.9) and (10.10) it follows that

$$\sum_{i=1}^{\infty} \sum_{i=1}^{M_n(S_n)} P(F_n^*) \geq \sum_{i=1}^{\infty} c_{95} \frac{(2^{-n-1} - 2^{-n})2^{-(n-1)N} \exp\left\{-\frac{1}{2}\varphi^2(2^{-n+1})\right\}}{\varepsilon_n^N \varphi(2^{-n+1})} \\ \geq c_{95} I_f(\sigma, \varphi) = +\infty.$$

Now we shall check the condition (b) of Lemma 3 (ii). Set

$$\gamma_i^{(m)} = E[(X(s_i^{(m)}) - X(s_0))(X(s_j^{(m)}) - X(s_0))]/\sigma(r_i^{(m)})\sigma(r_j^{(m)}), \\ S_{m,k}^{(1)} = \{s_j^{(m)}; 1 - k/n \leq \gamma_j^{(m)} \leq 1 - (k-1)/n\}, \quad 1 \leq k \leq \sqrt{\log n} + 1, \\ S_m^{(2)} = \{s_j^{(m)}; 1/\varphi(r_i^{(m)})\varphi(r_j^{(m)}) \leq \gamma_j^{(m)} \leq 1 - 1/\sqrt{\log n}\} \\ m_0 = n + 3 \log_2 n / (\alpha \log 2),$$

for fixed n, i such that $r_i^{(n)} \geq r_j^{(m)}$. Since by the concavity of $\sigma^2(x)$ we see

$$(10.11) \quad \gamma_j^{(m)} = \frac{\sigma^2(r_i^{(n)}) + \sigma^2(r_j^{(m)}) - \sigma^2(r_i^{(n)} - r_j^{(m)})}{2\sigma(r_i^{(n)})\sigma(r_j^{(m)})} \\ \leq \sigma(r_j^{(m)})/\sigma(r_i^{(n)}),$$

it follows from (10.1) that

$$(10.12) \quad \gamma_j^{(m)} \varphi(r_i^{(n)}) \varphi(r_j^{(m)}) \leq \sigma(2^{-m-2}) \varphi(2^{-n-2}) \varphi(2^{-m+1}) / \sigma(2^{-n-1}) \\ \ll 1.$$

Using Lemma 4, (ii) we have

$$(10.13) \quad P(F_1^n \cap F_2^m) \leq d_3 P(F_1^n) P(F_2^m)$$

for $m \geq m_0$. While there exists a constant c_{97} such that

$$(10.14) \quad \begin{aligned} \gamma_j^{(m)} &\leq \sigma(r_j^{(m)}) / \sigma(r_i^{(n)}) \\ &\leq \sigma(2^{-n-c_{97}+2}) / \sigma(2^{-n+1}) \ll 1/2 \end{aligned}$$

and from concavity of $\sigma^2(x)$ we have

$$(10.15) \quad \begin{aligned} \gamma_j^{(m)} &\leq 1 - \frac{\sigma^2(r_i^{(n)} - r_j^{(m)}) - \sigma^2(r_j^{(m)}) (\sqrt{1 + \sigma^2(r_i^{(n)} - r_j^{(m)}) / \sigma^2(r_j^{(m)})} - 1)^2}{2\sigma(r_i^{(n)})\sigma(r_j^{(m)})} \\ &\leq 1 - c_{98} \sigma^2(r_i^{(n)} - r_j^{(m)}) / \sigma^2(2^{-n+2}) \quad \text{for } n \leq m \leq n + c_{97}. \end{aligned}$$

This yields

$$(10.16) \quad \begin{aligned} \# S_{m,n}^{(1)} &\leq c_{99} \left(\sigma^{-1} \left(\sqrt{\frac{k}{c_{98}n}} \sigma(2^{-n+2}) \right) / \epsilon_m \right)^N \\ &\leq c_{100} k^{nN} \quad \text{for } n \leq m \leq n + c_{97}. \end{aligned}$$

Combining (10.14), (10.15) and (10.16) we have

$$(10.17) \quad \begin{aligned} &\left(\sum_{m=n}^{n+c_{97}} \sum_{k=1}^{\sqrt{\log n+1}} \sum_{j \in S_{m,n}^{(1)}} + \sum_{m=n}^{m_1} \sum_{j \in S_{m,n}^{(2)}} \right) P(F_1^n \cap F_2^m) \\ &\leq \left[c_{97} \sum_{k=1}^{\infty} c_{100} k^{nN} \exp\{-c_{101}k\} + \frac{3 \log_{(2)} n}{\alpha \log 2} (\log n)^{N/\alpha} \right. \\ &\quad \left. \times \exp\{-c_{102}\sqrt{\log n}\} \right] P(F_1^n) \\ &\leq d_{26} P(F_1^n). \end{aligned}$$

This yields the proof of Theorem 7.

The proofs of Corollaries 5, 6 and Theorem 8 are given by the analogous method for Corollaries 3, 4 and Theorem 5. So we omit them.

11. Proofs of Propositions 1~4

It is obvious from Lemma 5 that if $\sigma(x)$ is a *n.r.v.f.*, there exists a constant c_{103} such that $\sigma(x) < c_{103} < +\infty$ and that if $\sigma(x)$ is a *n.s.v.f.* we get

$$(11.1) \quad \lim_{\epsilon \downarrow 0} F_{\sigma(x)} / \sigma(x) \geq \int_0^{\infty} \lim_{\epsilon \downarrow 0} \frac{\sigma(xe^{-u^2})}{\sigma(x)} du = +\infty$$

by Fatou's Lemma. This shows Proposition 1.

To prove Propositions 2 and 3 the following Lemma is useful.

Lemma 11. *Assume that $\sigma(x)$ is a n.s.v.f. with a structure function $a(x)$ satisfying the conditions (ii) (b) of Theorem 5 and (ii) of Proposition 2. Then*

$$(11.2) \quad \sigma(x) = F_\sigma(x)/\sigma(x) \leq c_{104}/\sqrt{a(x)}.$$

Proof. By the expression (2.11) of a s.v.f. we get

$$(11.3) \quad \begin{aligned} \sigma(x) &\leq c_{105} \int_0^\infty \exp\left\{-\int_{xe^{-s}}^{x^2} a(s)/s ds\right\} du \\ &= c_{106} \sqrt{\log 1/x} \int_0^\infty \exp\left\{-(\log 1/x) \int_1^{1+y^2} a(e^{-s \log 1/x}) ds\right\} dy. \end{aligned}$$

Set

$$(11.4) \quad f_x(y) = \int_1^{1+y^2} a(e^{-s \log 1/x}) ds.$$

For a fixed x , $f_x(y)$ is a strictly increasing function of y with continuous positive derivative and so the inverse function $f_x^{-1}(z)$ has the same properties. In (11.3) changing the variable y to $z = (\log 1/x)f_x(y)$, (11.3) yields

$$(11.5) \quad \begin{aligned} &\sqrt{\log 1/x} \int_0^\infty \exp\left\{-\log(1/x) f_x(y)\right\} dy \\ &= \sqrt{\log 1/x} \int_0^\infty e^{-z} df_x^{-1}(z/\log 1/x). \end{aligned}$$

Since by the assumption (ii) (b) of Theorem 5 we get

$$(11.6) \quad z = (\log 1/x) f_x(y) \gg \int_1^{1+y^2} \gamma/s ds = \gamma \log(1+y^2) \quad (x \downarrow 0),$$

it holds that

$$(11.7) \quad f_x^{-1}(z/\log 1/x) \ll \exp(z/2\gamma), \quad (x \downarrow 0).$$

Therefore we have

$$(11.8) \quad \int_0^\infty e^{-z} f_x^{-1}(z/\log 1/x) dz < +\infty.$$

Integrating by part the right hand of (11.5), we get

$$\begin{aligned}
 (11.9) \quad & \sqrt{\log 1/x} \int_0^{\infty} e^{-z} df_x^{-1}(z/\log 1/x) \\
 &= \sqrt{\log 1/x} \int_0^{\infty} e^{-z} f_x^{-1}(z/\log 1/x) dz \\
 &= \sqrt{\log 1/x} \int_0^{(\log 1/x) f_x(1)} e^{-z} f_x^{-1}(z/\log 1/x) dz \\
 &\quad + \sqrt{\log 1/x} \int_{(\log 1/x) f_x(1)}^{\infty} e^{-z} f_x^{-1}(z/\log 1/x) dz \\
 &= I_1 + I_2.
 \end{aligned}$$

To estimate I_1 , set $f(z/(\log 1/x))=y$. If z changes a value 0 to $(\log 1/x)f_x(1)$ then y takes a value 0 to 1. Then by the assumption (ii) of Proposition 2 we have

$$(11.10) \quad \frac{z}{\log 1/x} = f_x(y) \geq c_{106} a(x) y^2 \quad \text{for } 0 \leq y \leq 1.$$

Hence we have

$$\begin{aligned}
 (11.11) \quad I_1 &\ll \sqrt{\log 1/x} \int_0^{(\log 1/x) f_x(1)} e^{-z} \frac{1}{\sqrt{c_{106} a(x)}} \cdot \frac{\sqrt{z}}{\sqrt{\log 1/x}} dz \\
 &\leq 2 \frac{\sqrt{\pi}}{\sqrt{c_{106} a(x)}}.
 \end{aligned}$$

Next by (11.7) and (11.10) we get

$$\begin{aligned}
 (11.12) \quad I_2 &\ll \sqrt{\log 1/x} \int_{(\log 1/x) c_{106} a(x)}^{\infty} e^{-z-z/2\gamma} dz \\
 &\leq \sqrt{\frac{2e}{c_{106}}} \left(\frac{2\gamma}{2\gamma-1} \right)^{3/2} \frac{1}{\sqrt{a(x)}}.
 \end{aligned}$$

Thus combining (11.9), (11.11) and (11.12) one gets the proof of Lemma 11. Proposition 2 is an immediate consequence of Lemma 11. If $\sigma(x)$ satisfies the condition of Proposition 3 we have

$$(11.13) \quad \sigma(x) \sqrt{\log 1/x} \leq c_{107}.$$

To estimate the lower bound of (11.13), it is sufficient to show the inequality

$$\begin{aligned}
 (11.14) \quad \sigma(x) &\geq G_\sigma(\sqrt{\log 1/x})/\sigma(x) \\
 &\gg c_{108} \sqrt{\log 1/x},
 \end{aligned}$$

under the assumption of Proposition 3.

In fact we see

$$\begin{aligned}
 (11.15) \quad & G_\sigma(\sqrt{\log 1/x})/\sigma(x) \\
 & \geq c_{108} \int_{\sqrt{\log 1/x}}^{\infty} \exp\left\{-\frac{c_4(y^2 - \log 1/x)}{\log 1/x}\right\} dy \\
 & = \frac{c_{108} e^{c_4}}{\sqrt{c_4}} \int_{c_4}^{\infty} e^{-y^2} dy \sqrt{\log 1/x}.
 \end{aligned}$$

This completes the proof of Proposition 3.

To prove Proposition 4 it is sufficient to show the following Lemma.

Lemma 12. *Assume that $\sigma(x)$ is a n.s.v.f. with a structure function $a(x)$ which satisfies the condition of Proposition 4. Then it follows that*

$$(11.16) \quad \sigma(x) \leq c_{110} \sqrt{\log 1/x} (\log_{(2)} 1/x) \cdots (\log_{(m)} 1/x),$$

$$(11.17) \quad G_\sigma(\sqrt{\log 1/x})/\sigma(x) \geq c_{111} \sqrt{\log 1/x} (\log_{(2)} 1/x) \cdots (\log_{(m)} 1/x).$$

Proof. Set

$$\begin{aligned}
 g(y) &= \sqrt{y} (\log y) \cdots (\log_{(m-2)} y) (\log_{(m-1)} y)^{c_1}, \\
 e(x) &= e(x), \quad e^m(x) = \exp(e^{m-1}(x)).
 \end{aligned}$$

It follows from (11.3) that

$$\begin{aligned}
 (11.18) \quad & \sigma(x) \leq c_{105} \sqrt{\log 1/x} \int_0^{\infty} \exp\left\{-(\log 1/x) \int_1^{1+y^2} a(e^{-s \log 1/x}) ds\right\} dy \\
 & = c_{105} \sqrt{\log 1/x} \int_0^1 \exp\left\{-\int_{\log 1/x}^{(y^2+1)\log 1/x} a(e^{-u}) du\right\} dy \\
 & \quad + c_{105} \sqrt{\log 1/x} \int_1^{\infty} \exp\left\{-\int_{\log 1/x}^{(y^2+1)\log 1/x} a(e^{-u}) du\right\} dy \\
 & = J_1 + J_2.
 \end{aligned}$$

Obviously we see

$$(11.19) \quad J_1 \leq c_{105} \sqrt{\log 1/x}.$$

By the condition of Proposition 4, we have

$$\begin{aligned}
 (11.20) \quad & J_2 \ll c_{105} \sqrt{\log 1/x} \int_1^{\infty} \exp\left\{-\int_{\log 1/x}^{y^2 \log 1/x} g'(u)/g(u) du\right\} dy \\
 & \leq c_{112} \sqrt{\log 1/x} (\log_{(2)} 1/x) \cdots (\log_{(m)} 1/x).
 \end{aligned}$$

This yields (11.16).

Finally we shall prove (11.17). It follows that

$$\begin{aligned}
 (11.21) \quad & G_\sigma(\sqrt{\log 1/x})/\sigma(x) \\
 & \geq c_{113} \int_{\sqrt{\log 1/x}}^{\infty} \frac{g(\log 1/x)}{g(y^2)} \exp\left\{-\int_{\log 1/x}^{y^2} \left(a(e^{-u}) - \frac{g'(u)}{g(u)}\right) du\right\} dy \\
 & = c_{113} g(\log 1/x) \sum_{n=1}^{\infty} \int_{a_n}^{a_{n+1}} \frac{\exp\left\{-\int_{\log 1/x}^{y^2} \left(a(e^{-u}) - \frac{g'(u)}{g(u)}\right) du\right\}}{g(y^2)} dy \\
 & \gg c_{114} g(\log 1/x) (\log_{(m)} 1/x)^{\epsilon_1 - \epsilon_2} \sum_{n=1}^{\infty} \int_{a_n}^{a_{n+1}} \frac{(\log_{(m-1)} y^2)^{\epsilon_1 - \epsilon_2}}{g(y)^2} dy \\
 & \geq c_{114} \sqrt{\log 1/x} (\log_{(2)} 1/x) \cdots (\log_{(m)} 1/x),
 \end{aligned}$$

where $a_n = e\left(\frac{1}{2} e^{m-3} ((\log_{(m-1)} 1/x)^n)\right)$.

This completes the proof of Lemma 12.

OSAKA UNIVERSITY

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