

SHARP INEQUALITY FOR BOUNDED SUBMARTINGALES AND THEIR DIFFERENTIAL SUBORDINATES

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ABSTRACT. For any fixed $\alpha \in [0, 1]$ and $\lambda > 0$ we determine the optimal function $V_{\alpha, \lambda}$ satisfying

$$\mathbb{P}(\max_n |g_n| \geq \lambda) \leq \mathbb{E}V_{\alpha, \lambda}(f_0, g_0)$$

for any submartingale $f = (f_n)$ bounded in absolute value by 1 and any process $g = (g_n)$ which is real-valued, adapted, integrable and satisfying

$$|dg_n| \leq |df_n| \text{ and } |\mathbb{E}(dg_n | \mathcal{F}_{n-1})| \leq \alpha \mathbb{E}(df_n | \mathcal{F}_{n-1}), \quad n = 1, 2, \dots,$$

with probability 1. As a corollary, a sharp exponential inequality for the distribution function of $\max_n |g_n|$ is established.

1. INTRODUCTION

Let $(\Omega, \mathcal{F}, \mathbb{P})$ be a probability space, equipped with a discrete filtration (\mathcal{F}_n) . Let $f = (f_n)_{n=0}^\infty$, $g = (g_n)_{n=0}^\infty$ be adapted integrable processes taking values in a certain separable Hilbert space \mathcal{H} . The difference sequences $df = (df_n)$, $dg = (dg_n)$ of these processes are given by

$$df_0 = f_0, \quad df_n = f_n - f_{n-1}, \quad dg_0 = g_0, \quad dg_n = g_n - g_{n-1}, \quad n = 1, 2, \dots$$

Let g^* stand for the maximal function of g , that is, $g^* = \max_n |g_n|$.

The following notion of differential subordination is due to Burkholder. The process g is differentially subordinate to f (or, in short, subordinate to f) if for any nonnegative integer n we have, almost surely,

$$|dg_n| \leq |df_n|.$$

We will slightly change this definition and say that g is differentially subordinate to f , if the above inequality for the differences holds for any *positive* integer n .

Let α be a fixed nonnegative number. Then g is α -differentially subordinate to f (or, in short, α -subordinate to f), if it is subordinate to f and for any positive integer n we have

$$|\mathbb{E}(dg_n | \mathcal{F}_{n-1})| \leq \alpha |\mathbb{E}(df_n | \mathcal{F}_{n-1})|.$$

This concept was introduced by Burkholder in [2] in the special case $\alpha = 1$. In general form, it first appeared in the paper by Choi [3].

In the sequel it will sometimes be convenient to work with simple processes. A process f is called simple, if for any n the variable f_n is simple and there exists N such that $f_N = f_{N+1} = f_{N+2} = \dots$. Given such a process, we will identify it with the finite sequence $(f_n)_{n=0}^N$.

Assume that the processes f and g are real-valued and fix $\alpha \in [0, 1]$. The objective of this paper is to establish a sharp exponential inequality for the distribution function of g^* under the assumption that f is a submartingale satisfying $\|f\|_\infty \leq 1$

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and g is α -subordinate to f . To be more precise, for any $\lambda > 0$ define the function $V_{\alpha,\lambda} : [-1, 1] \times \mathbb{R} \rightarrow \mathbb{R}$ by the formula

$$(1.1) \quad V_{\alpha,\lambda}(x_0, y_0) = \sup \mathbb{P}(g^* \geq \lambda).$$

Here the supremum is taken over all pairs (f, g) of integrable adapted processes, such that $(f_0, g_0) \equiv (x_0, y_0)$ almost surely, f is a submartingale satisfying $\|f\|_\infty \leq 1$ and g is α -subordinate to f . The filtration must also vary, as well as the probability space, unless it is nonatomic. Our main result is an explicit formula for the functions $V_{\alpha,\lambda}$, $\lambda > 0$. Usually we will omit the index α and write V_λ instead of $V_{\alpha,\lambda}$.

Let us discuss some related results which appeared in the literature. In [1] Burkholder studied the analogous question in the case of f, g being Hilbert space-valued martingales. The paper [1] contains also a related one-sided sharp exponential inequality for real martingales. This work was later extended by Hammack [4], who established a similar (two-sided) inequality under the assumption that f is a submartingale bounded by 1 and g is R^ν -valued, $\nu \geq 1$, and strongly 1-subordinate to f . Both papers present applications to stochastic integrals.

The paper is organized as follows. In the next section we introduce a family of special functions U_λ , $\lambda > 0$ and study their properties. This enables us to establish the inequality $V_\lambda \leq U_\lambda$ in Section 3. Then we prove the reverse inequality in the last section.

Throughout the paper, α is a fixed number from the interval $[0, 1]$. All the considered processes are assumed to be real valued.

2. THE EXPLICIT FORMULAS

Let S be the strip $[-1, 1] \times \mathbb{R}$. Consider the following subsets of S : for $0 < \lambda \leq 2$,

$$\begin{aligned} A_\lambda &= \{(x, y) \in S : |y| \geq x + \lambda - 1\}, \\ B_\lambda &= \{(x, y) \in S : 1 - x \leq |y| < x + \lambda - 1\}, \\ C_\lambda &= \{(x, y) \in S : |y| < 1 - x \text{ and } |y| < x + \lambda - 1\}. \end{aligned}$$

For $\lambda \in (2, 4)$, define

$$\begin{aligned} A_\lambda &= \{(x, y) \in S : |y| \geq \alpha x + \lambda - \alpha\}, \\ B_\lambda &= \{(x, y) \in S : \alpha x + \lambda - \alpha > |y| \geq x - 1 + \lambda\}, \\ C_\lambda &= \{(x, y) \in S : x - 1 + \lambda > |y| \geq 1 - x\}, \\ D_\lambda &= \{(x, y) \in S : 1 - x > |y| \geq -x - 3 + \lambda \text{ and } |y| < x - 1 + \lambda\}, \\ E_\lambda &= \{(x, y) \in S : -x - 3 + \lambda > |y|\}. \end{aligned}$$

Finally, for $\lambda \geq 4$, let

$$\begin{aligned} A_\lambda &= \{(x, y) \in S : |y| \geq \alpha x + \lambda - \alpha\}, \\ B_\lambda &= \{(x, y) \in S : \alpha x + \lambda - \alpha > |y| \geq x - 1 + \lambda\}, \\ C_\lambda &= \{(x, y) \in S : x - 1 + \lambda > |y| \geq -x - 3 + \lambda\}, \\ D_\lambda &= \{(x, y) \in S : -x - 3 + \lambda > |y| \geq 1 - x\}, \\ E_\lambda &= \{(x, y) \in S : 1 - x > |y|\}. \end{aligned}$$

Let $H : S \times (-1, \infty) \rightarrow \mathbb{R}$ be a function given by

$$(2.1) \quad H(x, y, z) = \frac{1}{\alpha + 2} \left[1 + \frac{(x + 1 + |y|)^{1/(\alpha+1)} ((\alpha + 1)(x + 1) - |y|)}{(1 + z)^{(\alpha+2)/(\alpha+1)}} \right].$$

Now we will define the special functions $U_\lambda : S \rightarrow \mathbb{R}$. For $0 < \lambda \leq 2$, let

$$(2.2) \quad U_\lambda(x, y) = \begin{cases} 1 & \text{if } (x, y) \in A_\lambda, \\ \frac{2-2x}{1+\lambda-x-|y|} & \text{if } (x, y) \in B_\lambda, \\ 1 - \frac{(\lambda-1+x-|y|)(\lambda-1+x+|y|)}{\lambda^2} & \text{if } (x, y) \in C_\lambda. \end{cases}$$

For $2 < \lambda < 4$, set

$$(2.3) \quad U_\lambda(x, y) = \begin{cases} 1 & \text{if } (x, y) \in A_\lambda, \\ 1 - (\alpha(x-1) - |y| + \lambda) \cdot \frac{2\lambda-4}{\lambda^2} & \text{if } (x, y) \in B_\lambda, \\ \frac{2-2x}{1+\lambda-x-|y|} - \frac{2(1-x)(1-\alpha)(\lambda-2)}{\lambda^2} & \text{if } (x, y) \in C_\lambda, \\ \frac{2(1-x)}{\lambda} \left[1 - \frac{(1-\alpha)(\lambda-2)}{\lambda} \right] - \frac{(1-x)^2 - |y|^2}{\lambda^2} & \text{if } (x, y) \in D_\lambda, \\ a_\lambda H(x, y, \lambda-3) + b_\lambda & \text{if } (x, y) \in E_\lambda, \end{cases}$$

where

$$(2.4) \quad a_\lambda = -\frac{2(1+\alpha)(\lambda-2)^2}{\lambda^2}, \quad b_\lambda = 1 - \frac{4(\lambda-2)(1-\alpha)}{\lambda^2}.$$

For $\lambda \geq 4$, set

$$(2.5) \quad U_\lambda(x, y) = \begin{cases} 1 & \text{if } (x, y) \in A_\lambda, \\ 1 - \frac{\alpha(x-1) - |y| + \lambda}{4} & \text{if } (x, y) \in B_\lambda, \\ \frac{2-2x}{1+\lambda-x-|y|} - \frac{(1-x)(1-\alpha)}{4} & \text{if } (x, y) \in C_\lambda, \\ \frac{(1-x)(1+\alpha)}{4} \exp\left(\frac{3+x+|y|-\lambda}{2(\alpha+1)}\right) & \text{if } (x, y) \in D_\lambda, \\ a_\lambda H(x, y, 1) + b_\lambda & \text{if } (x, y) \in E_\lambda, \end{cases}$$

where

$$(2.6) \quad a_\lambda = -b_\lambda = -\frac{(1+\alpha)}{2} \exp\left(\frac{4-\lambda}{2\alpha+2}\right).$$

For $\alpha = 1$, the formulas (2.2), (2.3), (2.5) give the special functions constructed by Hammack [4]. The key properties of U_λ are described in the two lemmas below.

Lemma 2.1. *For $\lambda > 2$, let $\phi_\lambda, \psi_\lambda$ denote the partial derivatives of U_λ with respect to x, y on the interiors of $A_\lambda, B_\lambda, C_\lambda, D_\lambda, E_\lambda$, extended continuously to the whole of these sets. The following statements hold.*

- (i) *The functions $U_\lambda, \lambda > 2$, are continuous on $S \setminus \{(1, \pm\lambda)\}$.*
- (ii) *Let*

$$S_\lambda = \{(x, y) \in [-1, 1] \times \mathbb{R} : |y| \neq \alpha x + \lambda - \alpha \text{ and } |y| \neq x + \lambda - 1\}.$$

Then

$$(2.7) \quad \phi_\lambda, \psi_\lambda, \lambda > 2, \text{ are continuous on } S_\lambda.$$

- (iii) *For any $(x, y) \in S$, the function $\lambda \mapsto U_\lambda(x, y), \lambda > 0$, is left-continuous.*
- (iv) *For any $\lambda > 2$ we have the inequality*

$$(2.8) \quad \phi_\lambda \leq -\alpha|\psi_\lambda|.$$

- (v) *For $\lambda > 2$ and any $(x, y) \in S$ we have $\chi_{\{|y| \geq \lambda\}} \leq U_\lambda(x, y) \leq 1$.*

Proof. We start with computing the derivatives. Let $y' = y/|y|$ stand for the sign of y , with $y' = 0$ if $y = 0$. For $\lambda \in (2, 4)$ we have

$$\phi_\lambda(x, y) = \begin{cases} 0 & \text{if } (x, y) \in A_\lambda, \\ -\frac{(2\lambda-4)\alpha}{\lambda^2} & \text{if } (x, y) \in B_\lambda, \\ -\frac{2\lambda-2|y|}{(1+\lambda-x-|y|)^2} + \frac{(2\lambda-4)(1-\alpha)}{\lambda^2} & \text{if } (x, y) \in C_\lambda, \\ -\frac{2}{\lambda} \left[1 - \frac{(1-\alpha)(\lambda-2)}{\lambda} \right] + \frac{2(1-x)}{\lambda^2} & \text{if } (x, y) \in D_\lambda, \\ -c_\lambda(x + |y| + 1)^{-\alpha/(\alpha+1)}(x + 1 + \frac{\alpha}{\alpha+1}|y|) & \text{if } (x, y) \in E_\lambda, \end{cases}$$

$$\psi_\lambda(x, y) = \begin{cases} 0 & \text{if } (x, y) \in A_\lambda, \\ \frac{2\lambda-4}{\lambda^2} y' & \text{if } (x, y) \in B_\lambda, \\ \frac{2-2x}{(1+\lambda-x-|y|)^2} y' & \text{if } (x, y) \in C_\lambda, \\ \frac{2y}{\lambda^2} & \text{if } (x, y) \in D_\lambda, \\ c_\lambda (x + |y| + 1)^{-\alpha/(\alpha+1)} \frac{y}{1+\alpha} & \text{if } (x, y) \in E_\lambda, \end{cases}$$

where

$$c_\lambda = 2(1 + \alpha)(\lambda - 2)^{\alpha/(\alpha+1)} \lambda^{-2}.$$

Finally, for $\lambda \geq 4$, set

$$\phi_\lambda(x, y) = \begin{cases} 0 & \text{if } (x, y) \in A_\lambda, \\ -\frac{\alpha}{4} & \text{if } (x, y) \in B_\lambda, \\ -\frac{2\lambda-2|y|}{(1+\lambda-x-|y|)^2} + \frac{1-\alpha}{4} & \text{if } (x, y) \in C_\lambda, \\ -\frac{x+1+2\alpha}{8} \exp\left(\frac{x+|y|+3-\lambda}{2(\alpha+1)}\right) & \text{if } (x, y) \in D_\lambda, \\ -c_\lambda (x + |y| + 1)^{-\alpha/(\alpha+1)} (x + 1 + \frac{\alpha}{\alpha+1}|y|) & \text{if } (x, y) \in E_\lambda, \end{cases}$$

$$\psi_\lambda(x, y) = \begin{cases} 0 & \text{if } (x, y) \in A_\lambda, \\ \frac{1}{4} y' & \text{if } (x, y) \in B_\lambda, \\ \frac{2-2x}{(1+\lambda-x-|y|)^2} y' & \text{if } (x, y) \in C_\lambda, \\ \frac{(1-x)}{8} \exp\left(\frac{x+|y|+3-\lambda}{2(\alpha+1)}\right) y' & \text{if } (x, y) \in D_\lambda, \\ c_\lambda (x + |y| + 1)^{-\alpha/(\alpha+1)} \frac{y}{1+\alpha} & \text{if } (x, y) \in E_\lambda, \end{cases}$$

where

$$c_\lambda = (1 + \alpha) 2^{-(2\alpha+3)/(\alpha+1)} \exp\left(\frac{4 - \lambda}{2(\alpha + 1)}\right).$$

Now the properties (i), (ii), (iii) follow by straightforward computation. To prove (iv), note first that for any $\lambda > 2$ the condition (2.8) is clearly satisfied on the sets A_λ and B_λ . Suppose $(x, y) \in C_\lambda$. Then $\lambda - |y| \in [0, 4]$, $1 - x \leq \min\{\lambda - |y|, 4 - \lambda + |y|\}$ and (2.8) takes form

$$-2(\lambda - |y|) + \frac{2\lambda - 4}{\lambda^2} (1 - \alpha)(1 - x + \lambda - |y|)^2 + 2\alpha(1 - x) \leq 0,$$

or

$$(2.9) \quad -2(\lambda - |y|) + \frac{1 - \alpha}{4} \cdot (1 - x + \lambda - |y|)^2 + 2\alpha(1 - x) \leq 0,$$

depending on whether $\lambda < 4$ or $\lambda \geq 4$. As $(2\lambda - 4)/\lambda^2 \leq \frac{1}{4}$, it suffices to show (2.9). If $\lambda - |y| \leq 2$, then, as $1 - x \leq \lambda - |y|$, the left-hand side does not exceed

$$\begin{aligned} -2(\lambda - |y|) + (1 - \alpha)(\lambda - |y|)^2 + 2\alpha(\lambda - |y|) &= (\lambda - |y|)(-2 + (1 - \alpha)(\lambda - |y|) + 2\alpha) \\ &\leq (\lambda - |y|)(-2 + 2(1 - \alpha) + 2\alpha) = 0. \end{aligned}$$

Similarly, if $\lambda - |y| \in (2, 4]$, then we use the bound $1 - x \leq 4 - \lambda + |y|$ and conclude that the left-hand side of (2.9) is not greater than

$$-2(\lambda - |y|) + 4(1 - \alpha) + 2\alpha(4 - \lambda + |y|) = -2(\lambda - |y| - 2)(1 + \alpha) \leq 0$$

and we are done with the case $(x, y) \in C_\lambda$.

Assume that $(x, y) \in D_\lambda$. For $\lambda \in (2, 4)$, the inequality (2.8) is equivalent to

$$-\frac{2}{\lambda} \left[1 - \frac{(1 - \alpha)(\lambda - 2)}{\lambda} \right] + \frac{2 - 2x}{\lambda^2} \leq -\frac{2\alpha|y|}{\lambda^2},$$

or, after some simplifications, $\alpha|y| + 1 - x \leq 2 + \alpha\lambda - 2\alpha$. It is easy to check that $\alpha|y| + 1 - x$ attains its maximum for $x = -1$ and $|y| = \lambda - 2$ and then we have the equality. If $(x, y) \in D_\lambda$ and $\lambda \geq 4$, then (2.8) takes form $-(2\alpha + 1 + x) \leq -\alpha(1 - x)$, or $(x + 1)(\alpha + 1) \geq 0$. Finally, on the set E_λ , the inequality (2.8) is obvious.

(v) By (2.8), we have $\phi_\lambda \leq 0$, so $U_\lambda(x, y) \geq U_\lambda(1, y) = \chi_{\{|y| \geq \lambda\}}$. Furthermore, as $U_\lambda(x, y) = 1$ for $|y| \geq \lambda$ and $\psi_\lambda(x, y)y' \geq 0$ on S_λ , the second estimate follows. \square

Lemma 2.2. *Let x, h, y, k be fixed real numbers, satisfying $x, x+h \in [-1, 1]$ and $|k| \leq |h|$. Then for any $\lambda > 2$ and $\alpha \in [0, 1)$,*

$$(2.10) \quad U_\lambda(x+h, y+k) \leq U_\lambda(x, y) + \phi_\lambda(x, y)h + \psi_\lambda(x, y)k.$$

We will need the following fact, proved by Burkholder; see page 17 of [1].

Lemma 2.3. *Let x, h, y, k, z be real numbers satisfying $|k| \leq |h|$ and $z > -1$. Then the function*

$$F(t) = H(x+th, y+tk, z),$$

defined on $\{t : |x+th| \leq 1\}$, is convex.

Proof of the Lemma 2.2. Consider the function

$$G(t) = G_{x,y,h,k}(t) = U_\lambda(x+th, y+tk),$$

defined on the set $\{t : |x+th| \leq 1\}$. It is easy to check that G is continuous. As explained in [1], the inequality (2.10) follows once the concavity of G is established. This will be done by proving the inequality $G'' \leq 0$ at the points, where G is twice differentiable and checking the inequality $G'_+(t) \leq G'_-(t)$ for those t , for which G is not differentiable (even once). Note that we may assume $t = 0$, by a translation argument $G''_{x,y,h,k}(t) = G''_{x+th,y+tk,h,k}(0)$, with analogous equalities for one-sided derivatives. Clearly, we may assume that $h \geq 0$, changing the signs of both h, k , if necessary. Due to the symmetry of U_λ , we are allowed to consider $y \geq 0$ only.

We start from the observation that $G''(0) = 0$ on the interior of A_λ and $G'_+(0) \leq G'_-(0)$ for $(x, y) \in A_\lambda \cap \overline{B}_\lambda$. The latter inequality holds since $U_\lambda \equiv 1$ on A_λ and $U_\lambda \leq 1$ on B_λ . For the remaining inequalities, we consider the cases $\lambda \in (2, 4)$, $\lambda \geq 4$ separately.

The case $\lambda \in (2, 4)$. The inequality $G''(0) \leq 0$ is clear for (x, y) lying in the interior of B_λ . On C_λ , we have

$$(2.11) \quad G''(0) = -\frac{4(h+k)(h(\lambda-y) - k(1-x))}{(1-x-y+\lambda)^3} \leq 0,$$

which follows from $|k| \leq h$ and the fact that $\lambda - y \geq 1 - x$. For (x, y) in the interior of D_λ ,

$$G''(0) = \frac{-h^2 + k^2}{\lambda^2} \leq 0,$$

as $|k| \leq h$. Finally, on E_λ , the concavity follows by Lemma 2.3.

It remains to check the inequalities for one-sided derivatives. By Lemma 2.1 (ii), the points (x, y) , for which G is not differentiable at 0, do not belong to S_λ . Since we excluded the set $A_\lambda \cap \overline{B}_\lambda$, they lie on the line $y = x - 1 + \lambda$. For such points (x, y) , the left derivative equals

$$G'_-(0) = -\frac{2\lambda - 4}{\lambda^2}(\alpha h - k),$$

while the right one is given by

$$G'_+(0) = \frac{-h+k}{2(\lambda-y)} + \frac{(2\lambda-4)(1-\alpha)h}{\lambda^2},$$

or

$$G'_+(0) = -\frac{2h}{\lambda} \left[1 - \frac{(1-\alpha)(\lambda-2)}{\lambda} \right] + \frac{2(1-x)h + 2yk}{\lambda^2},$$

depending on whether $y \geq 1 - x$ or $y < 1 - x$. In the first case, the inequality $G'_+(0) \leq G'_-(0)$ reduces to

$$(h - k) \left(\frac{1}{2(\lambda - y)} - \frac{2(\lambda - 2)}{\lambda^2} \right) \geq 0,$$

while in the remaining one,

$$\frac{2}{\lambda^2} (h - k)(y - (\lambda - 2)) \geq 0.$$

Both inequalities follow from the estimate $\lambda - y \leq 2$ and the condition $|k| \leq h$.

The case $\lambda \geq 4$. On the set B_λ the concavity is clear. For C_λ , we have that the formula (2.11) holds. If (x, y) lies in the interior of D_λ , then

$$G''(0) = \frac{1}{8} \exp \left(\frac{3 + x + y - \lambda}{2(\alpha + 1)} \right) \left[\frac{1 - x}{2(\alpha + 1)} \cdot (-h^2 + k^2) - \left(2 - \frac{1 - x}{\alpha + 1} \right) (h^2 + hk) \right] \leq 0,$$

since $|k| \leq h$ and $(1 - x)/(\alpha + 1) \leq 2$. The concavity on E_λ is a consequence of Lemma 2.3. It remains to check the inequality for one-sided derivatives. By Lemma 2.1 (ii), we may assume $y = x + \lambda - 1$, and the inequality $G'_+(0) \leq G'_-(0)$ reads

$$\frac{1}{2} (h - k) \left(\frac{1}{\lambda - y} - \frac{1}{2} \right) \geq 0,$$

an obvious one, as $\lambda - y \leq 2$. □

3. THE MAIN THEOREM

Now we may state and prove the main result of the paper.

Theorem 3.1. *Suppose f is a submartingale satisfying $\|f\|_\infty \leq 1$ and g is an adapted process which is α -subordinate to f . Then for all $\lambda > 0$ we have*

$$(3.1) \quad \mathbb{P}(g^* \geq \lambda) \leq \mathbb{E}U_\lambda(f_0, g_0).$$

Proof. If $\lambda \leq 2$, then this follows immediately from the result of Hammack [4]; indeed, note that U_λ coincides with Hammack's special function and, furthermore, since g is α -subordinate to f , it is also 1-subordinate to f .

Fix $\lambda > 2$. We may assume $\alpha < 1$. It suffices to show that for any nonnegative integer n ,

$$(3.2) \quad \mathbb{P}(|g_n| \geq \lambda) \leq \mathbb{E}U_\lambda(f_0, g_0).$$

To see that this implies (3.1), fix $\varepsilon > 0$ and consider a stopping time $\tau = \inf\{k : |g_k| \geq \lambda - \varepsilon\}$. The process $f^\tau = (f_{\tau \wedge n})$, by Doob's optional sampling theorem, is a submartingale. Furthermore, we obviously have that $\|f^\tau\|_\infty \leq 1$ and the process $g^\tau = (g_{\tau \wedge n})$ is α -subordinate to f^τ . Therefore, by (3.2),

$$\mathbb{P}(|g_n^\tau| \geq \lambda - \varepsilon) \leq \mathbb{E}U_{\lambda - \varepsilon}(f_0^\tau, g_0^\tau) = \mathbb{E}U_{\lambda - \varepsilon}(f_0, g_0).$$

Now if we let $n \rightarrow \infty$, we obtain $\mathbb{P}(g^* \geq \lambda) \leq \mathbb{E}U_{\lambda - \varepsilon}(f_0, g_0)$ and by left-continuity of U_λ as a function of λ , (3.1) follows.

Thus it remains to establish (3.2). By Lemma 2.1 (v), $\mathbb{P}(|g_n| \geq \lambda) \leq \mathbb{E}U_\lambda(f_n, g_n)$ and it suffices to show that for all $1 \leq j \leq n$ we have

$$(3.3) \quad \mathbb{E}U_\lambda(f_j, g_j) \leq \mathbb{E}U_\lambda(f_{j-1}, g_{j-1}).$$

To do this, note that, since $|dg_j| \leq |df_j|$ almost surely, the inequality (2.10) yields

$$(3.4) \quad U_\lambda(f_j, g_j) \leq U_\lambda(f_{j-1}, g_{j-1}) + \phi_\lambda(f_{j-1}, g_{j-1})df_j + \psi_\lambda(f_{j-1}, g_{j-1})dg_j$$

with probability 1. Assume for now that $\phi_\lambda(f_{j-1}, g_{j-1})df_j$, $\psi_\lambda(f_{j-1}, g_{j-1})dg_j$ are integrable. By α -subordination, the condition (2.8) and the submartingale property $\mathbb{E}(d_j|\mathcal{F}_{j-1}) \geq 0$, we have

$$\begin{aligned} & \mathbb{E}[\phi_\lambda(f_{j-1}, g_{j-1})df_j + \psi_\lambda(f_{j-1}, g_{j-1})dg_j|\mathcal{F}_{j-1}] \\ & \leq \phi_\lambda(f_{j-1}, g_{j-1})\mathbb{E}(df_j|\mathcal{F}_{j-1}) + |\psi_\lambda(f_{j-1}, g_{j-1})| \cdot |\mathbb{E}(dg_j|\mathcal{F}_{j-1})| \\ & \leq [\phi_\lambda(f_{j-1}, g_{j-1}) + \alpha|\psi_\lambda(f_{j-1}, g_{j-1})|]\mathbb{E}(df_j|\mathcal{F}_{j-1}) \leq 0. \end{aligned}$$

Therefore, it suffices to take the expectation of both sides of (3.4) to obtain (3.3).

Thus the proof will be complete if we show the integrability of $\phi_\lambda(f_{j-1}, g_{j-1})df_j$ and $\psi_\lambda(f_{j-1}, g_{j-1})dg_j$. In both the cases $\lambda \in (2, 4)$, $\lambda \geq 4$, all we need is that the variables

$$(3.5) \quad \frac{2\lambda - 2|g_{j-1}|}{(1 - f_{j-1} - |g_{j-1}| + \lambda)^2}df_j \quad \text{and} \quad \frac{2 - 2f_{j-1}}{(1 - f_{j-1} - |g_{j-1}| + \lambda)^2}dg_j$$

are integrable on the set $K = \{|g_{j-1}| < f_{j-1} + \lambda - 1, |g_{j-1}| \geq \lambda - 1\}$, since outside it the derivatives ϕ_λ , ψ_λ are bounded by a constant depending only on α , λ and $|df_j|$, $|dg_j|$ do not exceed 2. The integrability is proved exactly in the same manner as in [4]. We omit the details. \square

We will now establish the following sharp exponential inequality.

Theorem 3.2. *Suppose f is a submartingale satisfying $\|f\|_\infty \leq 1$ and g is an adapted process which is α -subordinate to f . In addition, assume that $|g_0| \leq |f_0|$ with probability 1. Then for $\lambda \geq 4$ we have*

$$(3.6) \quad \mathbb{P}(g^* \geq \lambda) \leq \gamma e^{-\lambda/(2\alpha+2)},$$

where

$$\gamma = \frac{1 + \alpha}{2\alpha + 4} \left(\alpha + 1 + 2^{-\frac{\alpha+2}{\alpha+1}} \right) \exp\left(\frac{2}{\alpha+1}\right).$$

The inequality is sharp.

This should be compared to Burkholder's estimate (Theorem 8.1 in [1])

$$\mathbb{P}(g^* \geq \lambda) \leq \frac{e^2}{4} \cdot e^{-\lambda}, \quad \lambda \geq 2,$$

in the case when f, g are Hilbert space-valued martingales and g is subordinate to f . For $\alpha = 1$, we obtain the inequality of Hammack [4],

$$\mathbb{P}(g^* \geq \lambda) \leq \frac{(8 + \sqrt{2})e}{12} \cdot e^{-\lambda/4}, \quad \lambda \geq 4.$$

Proof of the inequality (3.6). We will prove that the maximum of U_λ on the set $K = \{(x, y) \in S : |y| \leq |x|\}$ is given by the right hand side of (3.6). This, together with the inequality (3.1) and the assumption $\mathbb{P}((f_0, g_0) \in K) = 1$, will imply the desired estimate. Clearly, by symmetry, we may restrict ourselves to the set $K^+ = K \cap \{y \geq 0\}$. If $(x, y) \in K^+$ and $x \geq 0$, then it is easy to check that

$$U_\lambda(x, y) \leq U_\lambda((x+y)/2, (x+y)/2).$$

Furthermore, a straightforward computation shows that the function $F : [0, 1] \rightarrow \mathbb{R}$ given by $F(s) = U_\lambda(s, s)$ is nonincreasing. Thus we have $U_\lambda(x, y) \leq U_\lambda(0, 0)$. On the other hand, if $(x, y) \in K^+$ and $x \leq 0$, then it is easy to prove that $U_\lambda(x, y) \leq U_\lambda(-1, x+y+1)$ and the function $G : [0, 1] \rightarrow \mathbb{R}$ given by $G(s) = U_\lambda(-1, s)$ is nondecreasing. Combining all these facts we have that for any $(x, y) \in K^+$,

$$(3.7) \quad U_\lambda(x, y) \leq U_\lambda(-1, 1) = \gamma e^{-\lambda/(2\alpha+2)}.$$

Thus (3.6) holds. The sharpness will be shown in the next section. \square

4. SHARPNESS

Recall the function $V_\lambda = V_{\alpha,\lambda}$ defined by (1.1) in the introduction. The main result in this section is Theorem 4.1 below, which, combined with Theorem 3.1, implies that the functions U_λ and V_λ coincide. If we apply this at the point $(-1, 1)$ and use the equality appearing in (3.7), we obtain that the inequality (3.6) is sharp.

Theorem 4.1. *For any $\lambda > 0$ we have*

$$(4.1) \quad U_\lambda \leq V_\lambda.$$

The main tool in the proof is the following „splicing” argument. Assume that the underlying probability space is the interval $[0, 1]$ with the Lebesgue measure.

Lemma 4.1. *Fix $(x_0, y_0) \in [-1, 1] \times \mathbb{R}$. Suppose there exists a filtration and a pair (f, g) of simple adapted processes, starting from (x_0, y_0) , such that f is a submartingale satisfying $\|f\|_\infty \leq 1$ and g is α -subordinate to f . Then $V_\lambda(x_0, y_0) \geq \mathbb{E}V_\lambda(f_\infty, g_\infty)$ for $\lambda > 0$.*

Proof. Let N be such that $(f_N, g_N) = (f_\infty, g_\infty)$ and fix $\varepsilon > 0$. With no loss of generality, we may assume that σ -field generated by f, g is generated by the family of intervals $\{[a_i, a_{i+1}) : i = 1, 2, \dots, M-1\}$, $0 = a_1 < a_2 < \dots < a_M = 1$. For any $i \in \{1, 2, \dots, M-1\}$, denote $x_0^i = f_N(a_i)$, $y_0^i = g_N(a_i)$. There exists a filtration and a pair (f^i, g^i) of adapted processes, with f being a submartingale bounded in absolute value by 1 and g being α -subordinate to f , which satisfy $f_0^i = x_0^i \chi_{[0,1)}$, $g_0^i = y_0^i \chi_{[0,1)}$ and $\mathbb{P}((g^i)^* \geq \lambda) > \mathbb{E}V_\lambda(f_0^i, g_0^i) - \varepsilon$. Define the processes F, G by $F_k = f_k, G_k = g_k$ if $k \leq N$ and

$$F_k(\omega) = \sum_{i=1}^{M-1} f_{k-N}^i((\omega - a_i)/(a_{i+1} - a_i)) \chi_{[a_i, a_{i+1})}(\omega),$$

$$G_k(\omega) = \sum_{i=1}^{M-1} g_{k-N}^i((\omega - a_i)/(a_{i+1} - a_i)) \chi_{[a_i, a_{i+1})}(\omega)$$

for $k > N$. It is easy to check that there exists a filtration, relative to which the process F is a submartingale satisfying $\|F\|_\infty \leq 1$ and G is an adapted process which is α -subordinate to F . Furthermore, we have

$$\begin{aligned} \mathbb{P}(G^* \geq \lambda) &\geq \sum_{i=1}^{M-1} (a_{i+1} - a_i) \mathbb{P}((g^i)^* \geq \lambda) \\ &> \sum_{i=1}^{M-1} (a_{i+1} - a_i) (\mathbb{E}V_\lambda(f_0^i, g_0^i) - \varepsilon) = \mathbb{E}V_\lambda(f_\infty, g_\infty) - \varepsilon. \end{aligned}$$

Since ε was arbitrary, the result follows. \square

Proof of Theorem 4.1. First note the following obvious properties of the functions $V_\lambda, \lambda > 0$: we have $V_\lambda \in [0, 1]$ and $V_\lambda(x, y) = V_\lambda(x, -y)$. The second equality is an immediate consequence of the fact that if g is α -subordinate to f , then so is $-g$.

In the proof of Theorem 4.1 we repeat several times the following procedure. Having fixed a point (x_0, y_0) from the strip S , we construct certain simple finite processes f, g starting from (x_0, y_0) , take their natural filtration (\mathcal{F}_n) , apply Lemma 4.1 and thus obtain a bound for $V_\lambda(x_0, y_0)$. All the constructed processes appearing in the proof below are easily checked to satisfy the conditions of this lemma: the condition $\|f\|_\infty \leq 1$ is straightforward, while the α -subordination and the fact that f is a submartingale are implied by the following. For any $n \geq 1$, either df_n satisfies $\mathbb{E}(df_n | \mathcal{F}_{n-1}) = 0$ and $dg_n = \pm df_n$, or $df_n \geq 0$ and $dg_n = \pm \alpha df_n$.

We will consider the cases $\lambda \leq 2$, $2 < \lambda < 4$, $\lambda \geq 4$ separately. Note that by symmetry, it suffices to establish (4.1) on $S \cap \{y \geq 0\}$.

The case $\lambda \leq 2$. Assume $(x_0, y_0) \in A_\lambda$. If $y_0 \geq \lambda$, then $g^* \geq \lambda$ almost surely, so $V_\lambda(x_0, y_0) \geq 1 = U_\lambda(x_0, y_0)$. If $\lambda > y_0 \geq \alpha x_0 - \alpha + \lambda$, then let $(f_0, g_0) \equiv (x_0, y_0)$,

$$(4.2) \quad df_1 = (1 - x_0)\chi_{[0,1]} \quad \text{and} \quad dg_1 = \alpha df_1.$$

Then we have $g_1 = y_0 + \alpha - \alpha x_0 \geq \lambda$, which implies $g^* \geq \lambda$ almost surely and (4.1) follows. Now suppose $(x_0, y_0) \in A_\lambda$ and $y_0 < \alpha x_0 - \alpha + \lambda$. Let $(f, g) \equiv (x_0, y_0)$,

$$(4.3) \quad df_1 = \frac{y_0 - x_0 + 1 - \lambda}{1 - \alpha} \chi_{[0,1]}, \quad dg_1 = \alpha df_1$$

and

$$(4.4) \quad df_2 = dg_2 = \beta \chi_{[0,1-\beta/2]} + (\beta - 2) \chi_{[1-\beta/2,1]},$$

where

$$(4.5) \quad \beta = \frac{\alpha x_0 - y_0 - \alpha + \lambda}{1 - \alpha} \in [0, 2].$$

Then (f_2, g_2) takes values $(-1, \lambda - 2)$, $(1, \lambda)$ with probabilities $\beta/2$, $1 - \beta/2$, respectively, so, by Lemma 4.1,

$$(4.6) \quad V_\lambda(x_0, y_0) \geq \frac{\beta}{2} V_\lambda(-1, \lambda - 2) + (1 - \frac{\beta}{2}) V_\lambda(1, \lambda) = \frac{\beta}{2} V_\lambda(-1, 2 - \lambda) + 1 - \frac{\beta}{2}.$$

Note that $(-1, 2 - \lambda) \in A_\lambda$. If $2 - \lambda \geq \alpha \cdot (-1) - \alpha + \lambda$, then, as already proved, $V_\lambda(-1, 2 - \lambda) = 1$ and $V_\lambda(x_0, y_0) \geq 1 = U_\lambda(x_0, y_0)$. If the converse inequality holds, i.e., $2 - \lambda < -2\alpha + \lambda$, then we may apply (4.6) to $x_0 = -1$, $y_0 = 2 - \lambda$ to get

$$V_\lambda(-1, 2 - \lambda) \geq \frac{\beta}{2} V_\lambda(-1, 2 - \lambda) + 1 - \frac{\beta}{2},$$

or $V_\lambda(-1, 2 - \lambda) \geq 1$. Thus we established $V_\lambda(x_0, y_0) = 1$ for any $(x_0, y_0) \in A_\lambda$.

Suppose then, that $(x_0, y_0) \in B_\lambda$. Let

$$(4.7) \quad \beta = \frac{2(1 - x_0)}{1 - x_0 - y_0 + \lambda} \in [0, 1]$$

and consider a pair (f, g) starting from (x_0, y_0) and satisfying

$$(4.8) \quad df_1 = -dg_1 = -\frac{x_0 - y_0 - 1 + \lambda}{2} \chi_{[0,\beta]} + (1 - x_0) \chi_{[\beta,1]}.$$

On $[0, \beta)$, the pair (f_1, g_1) lies in A_λ ; Lemma 4.1 implies $V_\lambda(x_0, y_0) \geq \beta = U_\lambda(x_0, y_0)$.

Finally, for $(x_0, y_0) \in C_\lambda$, let (f, g) start from (x_0, y_0) and

$$df_1 = -dg_1 = \frac{-x_0 - \lambda + 1 + y_0}{2} \chi_{[0,\gamma]} + \frac{y_0 - x_0 + 1}{2} \chi_{[\gamma,1]},$$

where

$$\gamma = \frac{y_0 - x_0 + 1}{\lambda} \in [0, 1].$$

On $[0, \gamma)$, the pair (f_1, g_1) lies in A_λ , while on $[\gamma, 1]$ we have $(f_1, g_1) = ((x_0 + y_0 + 1)/2, (x_0 + y_0 - 1)/2) \in B_\lambda$. Hence

$$V_\lambda(x_0, y_0) \geq \gamma \cdot 1 + (1 - \gamma) \cdot \frac{1 - x_0 - y_0}{\lambda} = U_\lambda(x_0, y_0).$$

The case $2 < \lambda < 4$. For $(x_0, y_0) \in A_\lambda$ we prove (4.1) using the same processes as in the previous case, i.e. the constant ones if $y_0 \geq \lambda$ and the ones given by (4.2) otherwise. The next step is to establish the inequality

$$(4.9) \quad V_\lambda(-1, \lambda - 2) \geq U_\lambda(-1, \lambda - 2) = \frac{1 + \alpha}{2} + \frac{1 - \alpha}{2} \cdot \left(\frac{4 - \lambda}{\lambda}\right)^2.$$

To do this, fix $\delta \in (0, 1]$ and set

$$\beta = \frac{\delta(1-\alpha)}{\lambda}, \quad \kappa = \frac{4-\lambda-\delta(1+\alpha)}{\lambda} \cdot \beta, \quad \gamma = \beta + (1-\beta) \cdot \frac{\delta(1+\alpha)}{4}, \quad \nu = \kappa \cdot \frac{\lambda}{4}.$$

We have $0 \leq \nu \leq \kappa \leq \beta \leq \gamma \leq 1$. Consider processes f, g given by $(f_0, g_0) \equiv (-1, \lambda - 2)$, $(df_1, dg_1) \equiv (\delta, \alpha\delta)$,

$$\begin{aligned} df_2 = -dg_2 &= \frac{\lambda - \delta(1-\alpha)}{2} \chi_{[0, \beta]} - \frac{\delta(1-\alpha)}{2} \chi_{[\beta, 1]}, \\ df_3 = dg_3 &= -\left(\lambda - 2 + \frac{\delta(1+\alpha)}{2}\right) \chi_{[0, \kappa]} + \left(2 - \frac{\lambda + \delta(1+\alpha)}{2}\right) \chi_{[\kappa, \beta]} \\ &\quad + \left(2 - \frac{\delta(1+\alpha)}{2}\right) \chi_{[\beta, \gamma]} - \frac{\delta(1+\alpha)}{2} \chi_{[\gamma, 1]}, \\ df_4 = -dg_4 &= \left(-2 + \frac{\lambda}{2}\right) \chi_{[0, \nu]} + \frac{\lambda}{2} \chi_{[\nu, \kappa]}. \end{aligned}$$

As $(f_4, |g_4|)$ takes values $(1, \lambda)$, $(1, 0)$ and $(-1, \lambda - 2)$ with probabilities $(\gamma - \beta) + (\kappa - \nu)$, $\beta - \kappa$ and $1 - \gamma + \nu$, respectively, we have

$$V_\lambda(-1, \lambda - 2) \geq \gamma - \beta + \kappa - \nu + (1 - \gamma + \nu)V_\lambda(-1, \lambda - 2),$$

or

$$V_\lambda(-1, \lambda - 2) \geq \frac{\gamma - \beta + \kappa - \nu}{\gamma - \nu} = \frac{1 + \alpha}{2} + \frac{1 - \alpha}{2} \cdot \left(\frac{4 - \lambda}{\lambda}\right)^2 - \frac{\delta(1 - \alpha^2)}{\lambda^2}.$$

As δ is arbitrary, we obtain (4.9). Now suppose $(x_0, y_0) \in B_\lambda$ and recall the pair (f, g) starting from (x_0, y_0) given by (4.3) and (4.4) (with β defined in (4.5)). As previously, it leads to (4.6), which takes form

$$\begin{aligned} V_\lambda(x_0, y_0) &\geq \frac{\beta}{2} \left[\frac{1 + \alpha}{2} + \frac{1 - \alpha}{2} \cdot \left(\frac{4 - \lambda}{\lambda}\right)^2 \right] + 1 - \frac{\beta}{2} \\ &= \frac{\beta(1 - \alpha)}{4} \left[\left(\frac{4 - \lambda}{\lambda}\right)^2 - 1 \right] + 1 = \frac{(\alpha x_0 - \alpha - y_0 + \lambda)(4 - 2\lambda)}{\lambda^2} + 1 = U_\lambda(x_0, y_0). \end{aligned}$$

For $(x_0, y_0) \in C_\lambda$, consider a pair (f, g) , starting from (x_0, y_0) defined by (4.8) (with β given by (4.7)). On $[0, \beta]$ we have $(f_1, g_1) = ((x_0 + y_0 + 1 - \lambda)/2, (x_0 + y_0 - 1 + \lambda)/2) \in B_\lambda$, so Lemma 4.1 yields

$$\begin{aligned} V_\lambda(x_0, y_0) &\geq \beta V_\lambda\left(\frac{x_0 + y_0 + 1 - \lambda}{2}, \frac{x_0 + y_0 - 1 + \lambda}{2}\right) \\ &= \frac{2(1 - x_0)}{1 + \lambda - x_0 - y_0} \cdot \left\{ 1 - \left[\alpha \left(\frac{x_0 + y_0 - 1 - \lambda}{2}\right) - \frac{x_0 + y_0 - 1 - \lambda}{2} \right] \cdot \frac{2\lambda - 4}{\lambda^2} \right\} \\ &= U_\lambda(x_0, y_0). \end{aligned}$$

For $(x_0, y_0) \in D_\lambda$, set $\beta = (y_0 - x_0 + 1)/\lambda \in [0, 1]$ and let a pair (f, g) be given by $(f_0, g_0) \equiv (x_0, y_0)$ and

$$df_1 = -dg_1 = \frac{-x_0 + y_0 + 1 - \lambda}{2} \chi_{[0, \beta]} + \frac{-x_0 + y_0 + 1}{2} \chi_{[\beta, 1]}.$$

As (f_1, g_1) takes values

$$\left(\frac{x_0 + y_0 + 1 - \lambda}{2}, \frac{x_0 + y_0 - 1 + \lambda}{2}\right) \in B_\lambda \text{ and } \left(\frac{x_0 + y_0 + 1}{2}, \frac{x_0 + y_0 - 1}{2}\right) \in C_\lambda$$

with probabilities β and $1 - \beta$, respectively, we obtain $V_\lambda(x_0, y_0)$ is not smaller than

$$\begin{aligned} &\beta V_\lambda\left(\frac{x_0 + y_0 + 1 - \lambda}{2}, \frac{x_0 + y_0 - 1 + \lambda}{2}\right) + (1 - \beta) V_\lambda\left(\frac{x_0 + y_0 + 1}{2}, \frac{x_0 + y_0 - 1}{2}\right) \\ &= \frac{y_0 - x_0 + 1}{\lambda} \cdot \left\{ 1 - \left[\alpha \left(\frac{x_0 + y_0 - 1 - \lambda}{2}\right) - \frac{x_0 + y_0 - 1 - \lambda}{2} \right] \cdot \frac{2\lambda - 4}{\lambda^2} \right\} \\ &\quad + \frac{\lambda - y_0 + x_0 - 1}{\lambda} \left[\frac{1 - x_0 - y_0}{\lambda} - \frac{(1 - x_0 - y_0)(1 - \alpha)(\lambda - 2)}{\lambda^2} \right] \end{aligned}$$

$$= I + II + III + IV,$$

where

$$I + III = \frac{y_0 - x_0 + 1}{\lambda} + \frac{(\lambda - y_0 + x_0 - 1)(1 - x_0 - y_0)}{\lambda^2} = \frac{2(1 - x_0)}{\lambda} - \frac{(1 - x_0)^2 - y_0^2}{\lambda^2}$$

and

$$\begin{aligned} II + IV &= \frac{(1 - \alpha)(\lambda - 2)}{\lambda^3} [(y_0 - x_0 + 1)(y_0 + x_0 - 1 - \lambda) - (1 - x_0 - y_0)(\lambda - y_0 + x_0 - 1)] \\ &= -\frac{(1 - \alpha)(\lambda - 2)}{\lambda^3} \cdot \lambda(2 - 2x_0). \end{aligned}$$

Combining these facts, we obtain $V_\lambda(x_0, y_0) \geq U_\lambda(x_0, y_0)$.

For $(x_0, y_0) \in E_\lambda$ with $(x_0, y_0) \neq (-1, 0)$, the following construction will turn to be useful. Denote $w = \lambda - 3$, so, as $(x_0, y_0) \in E_\lambda$, we have $x_0 + y_0 < w$. Fix positive integer N and set $\delta = \delta_N = (w - x_0 - y_0)/[N(\alpha + 1)]$. Consider sequences $(x_j^N)_{j=1}^{N+1}$, $(p_j^N)_{j=1}^{N+1}$, defined by

$$x_j^N = x_0 + y_0 + (j - 1)\delta(\alpha + 1), \quad j = 1, 2, \dots, N + 1,$$

and $p_1^N = (1 + x_0)/(1 + x_0 + y_0)$,

$$(4.10) \quad p_{j+1}^N = \frac{(1 + x_j^N)(1 + x_j^N + \frac{\delta(\alpha-1)}{2})p_j^N}{(1 + x_{j+1}^N)(1 + x_j^N + \frac{\delta(\alpha+1)}{2})} + \frac{\delta}{1 + x_{j+1}^N}, \quad j = 1, 2, \dots, N.$$

We construct a process (f, g) starting from (x_0, y_0) such that for $j = 1, 2, \dots, N + 1$,

$$(4.11) \quad \begin{aligned} &\text{the variable } (f_{3j}, |g_{3j}|) \text{ takes values } (x_j^N, 0) \text{ and } (-1, 1 + x_j^N) \\ &\text{with probabilities } p_j^N \text{ and } 1 - p_j^N, \text{ respectively.} \end{aligned}$$

We do this by induction. Let

$$df_1 = -dg_1 = y_0\chi_{[0, p_1^N]} + (-1 - x_0)\chi_{[p_1^N, 1]}, \quad df_2 = dg_2 = df_3 = dg_3 = 0.$$

Note that (4.11) is satisfied for $j = 1$. Now suppose we have a pair (f, g) , which satisfies (4.11) for $j = 1, 2, \dots, n$, $n \leq N$. Let us describe f_k and g_k for $k = 3n + 1, 3n + 2, 3n + 3$. The difference df_{3n+1} is determined by the following three conditions: it is a martingale difference, i.e., satisfies $\mathbb{E}(df_{3n+1}|\mathcal{F}_{3n}) = 0$; conditionally on $\{f_{3n} = x_n^N\}$, it takes values in $\{-1 - x_n^N, \delta(\alpha + 1)/2\}$; and vanishes on $\{f_{3n} \neq x_n^N\}$. Furthermore, set $dg_{3n+1} = df_{3n+1}$. Moreover,

$$df_{3n+2} = \delta\chi_{\{f_{3n+1} = -1\}}, \quad dg_{3n+2} = \frac{g_{3n+1}}{|g_{3n+1}|} \alpha \cdot df_{3n+2}.$$

Finally, the variable df_{3n+3} satisfies $\mathbb{E}(df_{3n+3}|\mathcal{F}_{3n+2}) = 0$, and, in addition, the variable f_{3n+3} takes values in $\{-1, x_n^N + \delta(\alpha + 1)\} = \{-1, x_n^{N+1}\}$. The description is completed by

$$dg_{3n+3} = -\frac{g_{3n+2}}{|g_{3n+2}|} df_{3n+3}.$$

One easily checks that $(f_{3n+3}, |g_{3n+3}|)$ takes values in $\{(x_{n+1}^N, 0), (-1, 1 + x_{n+1}^N)\}$; moreover, since

$$\begin{aligned} \mathbb{E}f_{3n+3} &= \mathbb{E}f_{3n} + \mathbb{E}df_{3n+2} = x_n^N p_n^N - (1 - p_n^N) + \delta\mathbb{P}(f_{3n+1} = -1) \\ &= x_n^N p_n^N - (1 - p_n^N) + \delta \left(1 - p_n^N + p_n^N \frac{\delta(\alpha + 1)}{2(1 + x_n^N) + \delta(\alpha + 1)} \right) \\ &= p_n^N \cdot \frac{(x_n^N + 1)(1 + x_n^N + \delta(\alpha - 1)/2)}{1 + x_n^N + \delta(\alpha + 1)/2} + \delta - 1, \end{aligned}$$

we see that $\mathbb{P}(f_{3n+3} = x_{n+1}^N) = p_{n+1}^N$ and the pair (f, g) satisfies (4.10) for $j = n + 1$. Thus there exists (f, g) satisfying (4.10) for $j = 1, 2, \dots, N + 1$. In particular,

$(f_{3(N+1)}, |g_{3(N+1)}|)$ takes values $(w, 0), (-1, w+1) \in D_\lambda$ with probabilities $p_{N+1}^N, 1 - p_{N+1}^N$. By Lemma 4.1,

$$(4.12) \quad V_\lambda(x_0, y_0) \geq p_{N+1}^N V_\lambda(w, 0) + (1 - p_{N+1}^N) V_\lambda(-1, w+1).$$

Recall the function H defined by (2.1). The function $h : [x_0 + y_0, w] \rightarrow \mathbb{R}$ given by $h(t) = H(x_0, y_0, t)$, satisfies the differential equation

$$h'(t) + \frac{\alpha + 2}{\alpha + 1} \cdot \frac{h(t)}{1 + t} = \frac{1}{(\alpha + 1)(1 + t)}.$$

As we assumed $x_0 + y_0 > -1$, the expression $(h(x+\delta) - h(x))/\delta$ converges uniformly to $h'(x)$ on $[x_0 + y_0, \lambda - 3]$. Therefore there exist constants ε_N , which depend only on N and $x_0 + y_0$ satisfying $\lim_{N \rightarrow \infty} \varepsilon_N = 0$ and for $1 \leq j \leq N$,

$$\left| \frac{h(x_{j+1}^N) - h(x_j^N)}{(\alpha + 1)\delta_N} + \frac{\left[\frac{\alpha+2}{\alpha+1}(1+x_j^N) - \frac{\delta_N(\alpha+1)}{2} \right] h(x_j^N)}{(1+x_{j+1}^N)(1+x_j^N + \frac{\delta_N(\alpha+1)}{2})} - \frac{1}{(\alpha+1)(1+x_{j+1}^N)} \right| \leq \varepsilon_N,$$

or, equivalently,

$$\left| h(x_{j+1}^N) - \frac{(1+x_j^N)(1+x_j^N + \frac{\delta_N(\alpha-1)}{2})h(x_j^N)}{(1+x_{j+1}^N)(1+x_j^N + \frac{\delta_N(\alpha+1)}{2})} - \frac{\delta_N}{1+x_{j+1}^N} \right| \leq (\alpha+1)\delta_N\varepsilon_N.$$

Together with (4.10), this leads to

$$|h(x_{j+1}^N) - p_{j+1}^N| \leq \frac{(1+x_j^N)(1+x_j^N + \frac{\delta_N(\alpha-1)}{2})}{(1+x_{j+1}^N)(1+x_j^N + \frac{\delta_N(\alpha+1)}{2})} |h(x_j^N) - p_j^N| + (\alpha+1)\delta_N\varepsilon_N.$$

Since $p_1^N = h(x_1^N)$, we have

$$|h(w) - p_{N+1}^N| \leq (\alpha+1)N\delta_N\varepsilon_N = (\lambda - 3 - x_0 - y_0)\varepsilon_N$$

and hence $\lim_{N \rightarrow \infty} p_{N+1}^N = h(w)$. Combining this with (4.12), we obtain

$$V_\lambda(x_0, y_0) \geq h(w)(V_\lambda(w, 0) - V_\lambda(-1, w+1)) + V_\lambda(-1, w+1).$$

As $w = \lambda - 3$, it suffices to check that we have

$$a_\lambda = V_\lambda(\lambda - 3, 0) - V_\lambda(-1, \lambda - 2) \text{ and } b_\lambda = V_\lambda(-1, \lambda - 2),$$

where a_λ, b_λ were defined in (2.4). Finally, if $(x_0, y_0) = (-1, 0)$, then considering a pair (f, g) starting from (x_0, y_0) and satisfying $df_1 \equiv \delta, dg_1 \equiv \alpha\delta$, we get

$$(4.13) \quad V(-1, 0) \geq V(-1 + \delta, \alpha\delta).$$

Now let $\delta \rightarrow 0$ to obtain $V(-1, 0) \geq U(-1, 0)$.

The case $\lambda \geq 4$. We proceed as in previous case. We deal with $(x_0, y_0) \in A_\lambda$ exactly in the same manner. Then we establish the analogue of (4.9), which is

$$(4.14) \quad V(-1, \lambda - 2) \geq U_\lambda(-1, \lambda - 2) = \frac{1 + \alpha}{2}.$$

To do this, fix $\delta \in (0, 1)$ and set

$$\beta = \frac{4 - 2\delta}{4 - \delta(1 + \alpha)}, \quad \gamma = \beta \cdot \left(1 - \frac{\delta(\alpha + 1)}{4} \right).$$

Now let a pair (f, g) be defined by $(f_0, g_0) \equiv (-1, \lambda - 2), (df_1, dg_1) \equiv (\delta, \alpha\delta)$,

$$df_2 = -dg_2 = -\frac{\delta(1 - \alpha)}{2}\chi_{[0, \beta)} + (2 - \delta)\chi_{[\beta, 1]},$$

$$df_3 = dg_3 = -\frac{\delta(1 + \alpha)}{2}\chi_{[0, \gamma)} + \left(2 - \frac{\delta(1 + \alpha)}{2} \right)\chi_{[\gamma, \beta)}.$$

Then (f_3, g_3) takes values $(-1, \lambda - 2)$, $(1, \lambda)$ and $(1, \lambda - 4 + \delta(\alpha + 1))$ with probabilities γ , $\beta - \gamma$ and $1 - \beta$, respectively, and Lemma 4.1 yields

$$V(-1, \lambda - 2) \geq \gamma V(-1, \lambda - 2) + (\beta - \gamma)V(1, \lambda),$$

or

$$V(-1, \lambda - 2) \geq \frac{\beta - \gamma}{1 - \gamma} = \frac{(\alpha + 1)(2 - \delta)}{4 - \delta(\alpha + 1)}.$$

It suffices to let $\delta \rightarrow 0$ to obtain (4.14). The cases $(x_0, y_0) \in B_\lambda$, C_λ are dealt with using the same processes as in the case $\lambda \in (2, 4)$. If $(x_0, y_0) \in D_\lambda$, then Lemma 4.1, applied to the pair (f, g) given by $(f_0, g_0) \equiv (x_0, y_0)$, $df_1 = -dg_1 = -(1 + x_0)\chi_{[0, (1-x_0)/2)} + (1 - x_0)\chi_{[(1-x_0)/2, 1]}$, yields

$$(4.15) \quad V(x_0, y_0) \geq \frac{1 - x_0}{2} V(-1, x_0 + y_0 + 1).$$

Furthermore, for any number y and any $\delta \in (0, 1)$, we have

$$(4.16) \quad V(-1, y) \geq V(-1 + \delta, y + \alpha\delta),$$

which is proved in the same manner as (4.13). Hence, for large N , if we set $\delta = (\lambda - 3 - x_0 - y_0)/(N(\alpha + 1))$, the inequalities (4.15) and (4.16) give

$$\begin{aligned} V(x_0, y_0) &\geq \frac{1 - x_0}{2} V(-1, x_0 + y_0 + 1) \geq \frac{1 - x_0}{2} V(-1 + \delta, x_0 + y_0 + 1 + \alpha\delta) \\ &\geq \frac{1 - x_0}{2} \left(1 - \frac{\delta}{2}\right) V(-1, x_0 + y_0 + 1 + (\alpha + 1)\delta) \\ &\geq \frac{1 - x_0}{2} \left(1 - \frac{\delta}{2}\right)^N V(-1, x_0 + y_0 + 1 + N(\alpha + 1)\delta) \\ &= \frac{1 - x_0}{2} \left(1 - \frac{\lambda - 3 - x_0 - y_0}{2N(\alpha + 1)}\right)^N V(-1, \lambda - 2) \\ &= \frac{(1 - x_0)(1 + \alpha)}{4} \left(1 - \frac{\lambda - 3 - x_0 - y_0}{2N(\alpha + 1)}\right)^N. \end{aligned}$$

Now take $N \rightarrow \infty$ to obtain $V_\lambda(x_0, y_0) \geq U_\lambda(x_0, y_0)$.

Finally, if $(x_0, y_0) \in E_\lambda$ we use the pair (f, g) used in the proof of the case $(x_0, y_0) \in E_\lambda$, $\lambda \in (2, 4)$, with $\omega = 1$. Then the process $(f, |g|)$ ends at the points $(1, 0)$ and $(-1, 2)$ with probabilities, which can be made arbitrarily close to $H(x_0, y_0, 1)$ and $1 - H(x_0, y_0, 1)$, respectively. It suffices to apply Lemma 4.1 and check that it gives $V_\lambda(x_0, y_0) \geq U_\lambda(x_0, y_0)$. \square

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