

# **Stochastic representations of max-type functionals from random walk**

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# Formulation of the problem

Consider the random walk

$$S_k = \xi_1 + \dots + \xi_k, \quad P(\xi_i = 1) = P(\xi_i = -1) = 1/2$$

on a probability space  $(\Omega, \mathcal{F}, P)$  with natural filtration  $\mathcal{F}_n = \sigma(\xi_i, i \leq n)$ ,  $\mathcal{F}_\infty = \sigma(\bigcup_{n \geq 0} \mathcal{F}_n)$ .

For functionals  $F = F(\omega)$  from random walk it required to make a

- **single representations**

$$F = c_F + \sum_k u_{k-1} \Delta S_k,$$

where  $\Delta S_k = S_k - S_{k-1}$ ,  $u_{k-1} \sim \mathcal{F}_{k-1}$  – measurable process.

- **multiple representations**

$$F = c_F + \sum_{k \geq 1} \sum_{t_1 < \dots < t_k} c_k(t_1, \dots, t_k) \Delta S_{t_1} \dots \Delta S_{t_k},$$

where  $c_k(t_1, \dots, t_k)$  is **deterministic** function.

There were considered the following functionals:

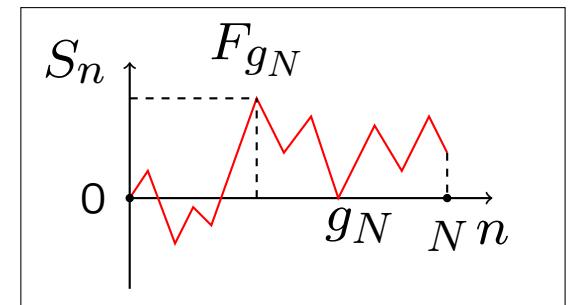
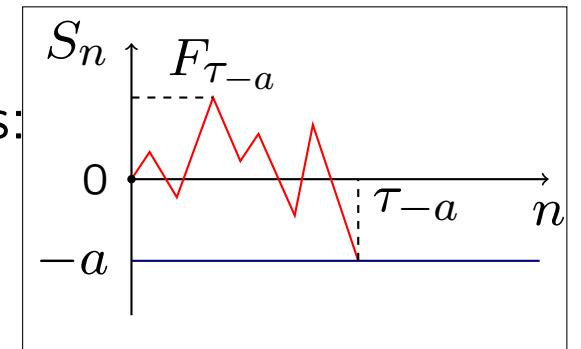
–  $F_N = \max_{0 \leq k \leq N} S_k;$

–  $F_{\tau-a} = \max_{0 \leq k \leq \tau-a} S_k,$

where  $\tau-a = \inf\{k \geq 0 : S_k = -a\}$ ,  $a \in \mathbb{N}$ .

–  $F_{g_N} = \max_{0 \leq k \leq g_N} S_k,$

where  $g_N = \sup\{0 < k \leq N : S_k = 0\}.$



## Single representations

1. Case  $F_N = \max_{0 \leq k \leq N} S_k$

**Theorem 1.** For  $F_N = \max_{0 \leq k \leq N} S_k$  we have the following stochastic representation:

$$\max_{0 \leq n \leq N} S_n = E(\max_{0 \leq n \leq N} S_n) + \sum_{k=1}^N G_{N-k+1}(F_{k-1} - S_{k-1}) \Delta S_k, \quad (1)$$

where

$$G_{N-k+1}(F_{k-1} - S_{k-1}) = \sum_{r=F_{k-1}-S_{k-1}+1}^{N-k+1} \sum_{l=0}^{\left[\frac{N-k+1-r}{2}\right]} (-1)^{\left[\frac{N-k+1-r}{2}\right]-l} \\ \times \binom{-1/2}{[(N-k+1-r)/2]-l} \varphi_{r+2l}(r),$$

$$\varphi_k(m) = \frac{m}{k2^k} \binom{k}{\frac{m+k}{2}}, \quad \mathbb{E}(\max_{0 \leq n \leq N} S_n) \sim \sqrt{\frac{2N}{\pi}} \quad \text{by} \quad N \rightarrow \infty.$$


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In continuous case for  $F_t = \max_{0 \leq s \leq t} B_s, t \leq T$

$$\boxed{\max_{0 \leq t \leq T} B_t = \mathbb{E}(\max_{0 \leq t \leq T} B_t) + 2 \int_0^T \left[ 1 - \Phi \left( \frac{F_t - B_t}{\sqrt{T-t}} \right) \right] dB_t}$$

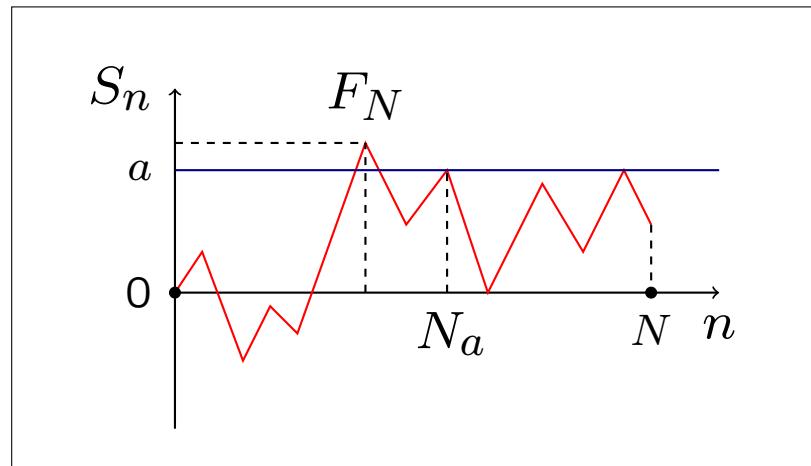
(**A. N. Shiryaev, M. Yor**)

Proof.

- $\xi \geq 0 \Rightarrow \xi = \int_0^{+\infty} I(\xi \geq a) da$ . Hence

$$F_N = \int_0^{+\infty} I(\max_{0 \leq k \leq N} S_k \geq a) da = \int_0^{+\infty} I(N_a \leq N) da,$$

where  $N_a = \inf\{k \geq 0 : S_k \geq a\}$ ,  $a \geq 0$ .



**Conclusion:** sufficiently to receive the representation for  $I(N_a \leq N)$ .

- **Discrete stochastic exponent** for  $\lambda \in (0, 1)$ :

$$\mathcal{E}_n(S) = \prod_{k=1}^n (1 + \lambda \Delta S_k) = (\sqrt{1 - \lambda^2})^n \left( \sqrt{\frac{1 + \lambda}{1 - \lambda}} \right)^{S_n}, \quad \mathcal{E}_0(S) := 1.$$

- $\Delta \mathcal{E}_n(S) = \lambda \mathcal{E}_{n-1}(S) \Delta S_n$  and  $\mathcal{E}_n(S)$  – martingale with respect to  $(\mathcal{F}_n)_{n \geq 0}$ .
- From the **optional sampling theorem** (P-a. s.)

$$\mathbb{E}(\mathcal{E}_{N_a} | \mathcal{F}_n) = \mathcal{E}_{N_a \wedge n} = 1 + \lambda \sum_{k=1}^{N_a \wedge n} \mathcal{E}_{k-1} \Delta S_k, \quad (2)$$

$$\mathbb{E} \mathcal{E}_{N_a} = 1 \Rightarrow \mathbb{E}(\sqrt{1 - \lambda^2})^{N_a} = \left( \sqrt{\frac{1 + \lambda}{1 - \lambda}} \right)^{-\lceil a \rceil}, \quad (3)$$

$$\left( \sqrt{\frac{1 + \lambda}{1 - \lambda}} \right)^{-\lceil a \rceil} = \sum_{l=0}^{+\infty} (\sqrt{1 - \lambda^2})^{\lceil a \rceil + 2l} \varphi_{\lceil a \rceil + 2l}(\lceil a \rceil). \quad (4)$$

From (2), (3), (4) we get

$$\begin{aligned} \mathbb{E}((\sqrt{1 - \lambda^2})^{N_a} | \mathcal{F}_n) &= \mathbb{E}(\sqrt{1 - \lambda^2})^{N_a} + \\ &+ \lambda \sum_{k=1}^{N_a \wedge n} \sum_{l=0}^{+\infty} (\sqrt{1 - \lambda^2})^{\lceil a \rceil - S_{k-1} + 2l + k - 1} \varphi_{\lceil a \rceil - S_{k-1} + 2l} (\lceil a \rceil - S_{k-1}) \Delta S_k, \end{aligned}$$

- Denote  $s := \sqrt{1 - \lambda^2}$ . Then for all  $s \in (0, 1)$  we have

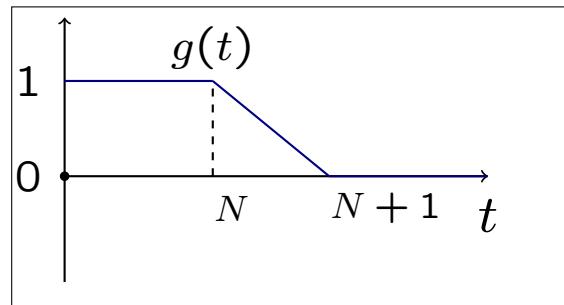
$$\begin{aligned} \mathbb{E}(s^{N_a} | \mathcal{F}_n) &= \mathbb{E}s^{N_a} + \sum_{k=1}^{N_a \wedge n} \sum_{l,m=0}^{+\infty} (-1)^m C_{\frac{1}{2}}^m \varphi_{\lceil a \rceil - S_{k-1} + 2l} (\lceil a \rceil - S_{k-1}) \\ &\quad \times s^{\lceil a \rceil - S_{k-1} + 2(m+l) + k - 1} \Delta S_k \quad (5) \end{aligned}$$

$$\left. \begin{array}{l} f_s(x) = s^x, x \geq 0, s \in (0, 1) \\ s = e^{-n}, n = 1, 2, \dots, \end{array} \right\} \Rightarrow (5) \text{ is valid for } \mathcal{L} = \{e^{-nx}\}_{n \geq 0}.$$

$$C_0[0, +\infty) = \{f \in C[0, +\infty) : f(x) \rightarrow 0 \text{ by } x \rightarrow +\infty\}$$

$\Downarrow_{(*)}$   
 $\mathcal{L} = \{e^{-nx}\}_{n \geq 0}$  **is complete in**  $C_0[0, +\infty)$ .

Consider the function



$$\left. \begin{array}{l} g(t) \in C_0[0, +\infty) \\ g(n) = I(n \leq N) \end{array} \right\} \Rightarrow (5) \text{ is valid for } I(n \leq N).$$

(\*) Sedletski A. M. *Classes of analytic Fourier transformations and their exponential approximations.*

Moscow: Physmatlit, 2005, 504 p.

**Lemma 1.** For  $I(N_a \leq N)$  we have the representation (P-a.s.)

$$I(N_a \leq N) = P(N_a \leq N) + \sum_{k=1}^{N_a \wedge N} \sum_{l,m: l+m \leq \left[\frac{N-k+1}{2}\right]} (-1)^m C_1^m \\ \times \varphi_{\lceil a \rceil - S_{k-1} + 2l}(\lceil a \rceil - S_{k-1}) I(\lceil a \rceil - S_{k-1} \leq N - k + 1 - 2(l+m)) \Delta S_k.$$

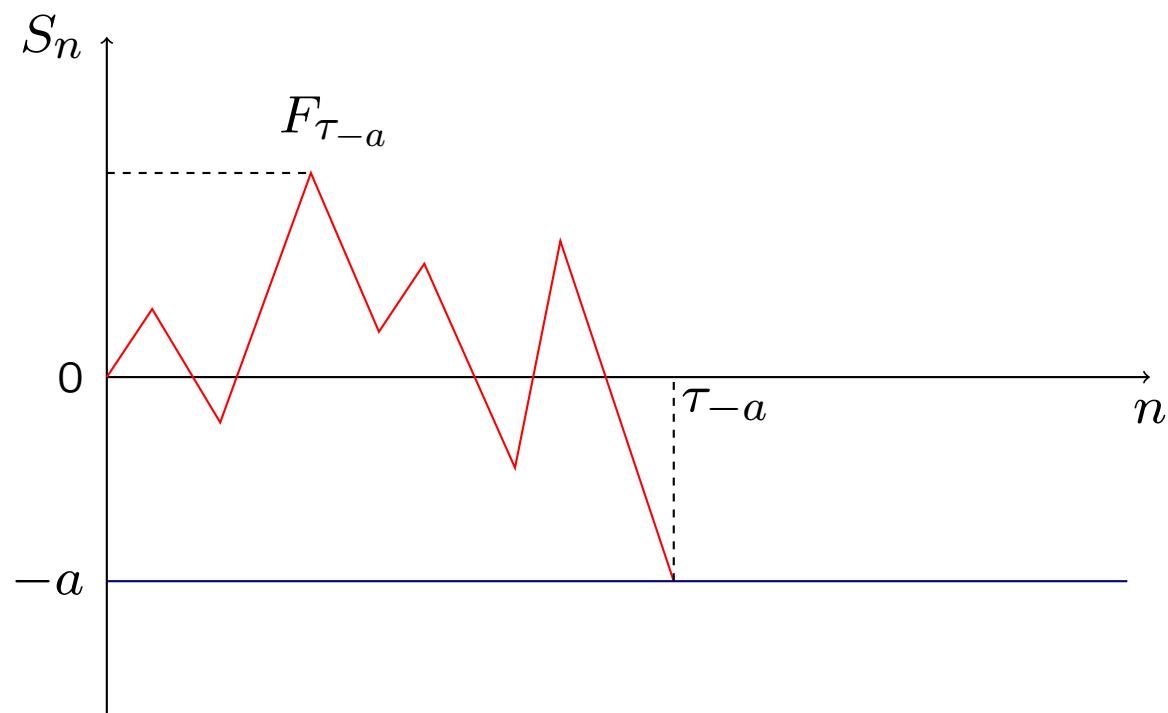
The decision formula is established by integrating on  $a$ .

$$F_N = E F_N + \sum_{k=1}^N G_{N-k+1}(F_{k-1} - S_{k-1}) \Delta S_k.$$

□

2. Case  $F_{\tau-a} = \max_{0 \leq k \leq \tau-a} S_k$ ,

$$\tau-a = \inf\{k \geq 0 : S_k = -a\}.$$



**Theorem 2.** For  $F_{\tau-a}$  we have a stochastic representation

$$\max_{0 \leq k \leq \tau-a} S_k = - \sum_{k=2}^{\tau-a} H(F_{k-1}) \Delta S_k,$$

where  $H(n) = \sum_{l=1}^n \frac{1}{a+l}$ ,  $F_k = \max_{0 \leq l \leq k} S_l$ .

Proof.

$$I(F_{\tau-a} \geq z) = I(N_z \leq \tau-a), \quad z > 0.$$



$$(a + \lceil z \rceil) I(F_{\tau-a} \geq z) = a + S_{\tau-a \wedge N_z}.$$

Then

$$S_{\tau-a \wedge N_z} + a = - \sum_{k=2}^{\tau-a} I(F_{k-1} \geq z) \Delta S_k,$$

$$I(F_{\tau-a} \geq z) = - \sum_{k=2}^{\tau-a} \frac{I(F_{k-1} \geq z)}{a + \lceil z \rceil} \Delta S_k.$$

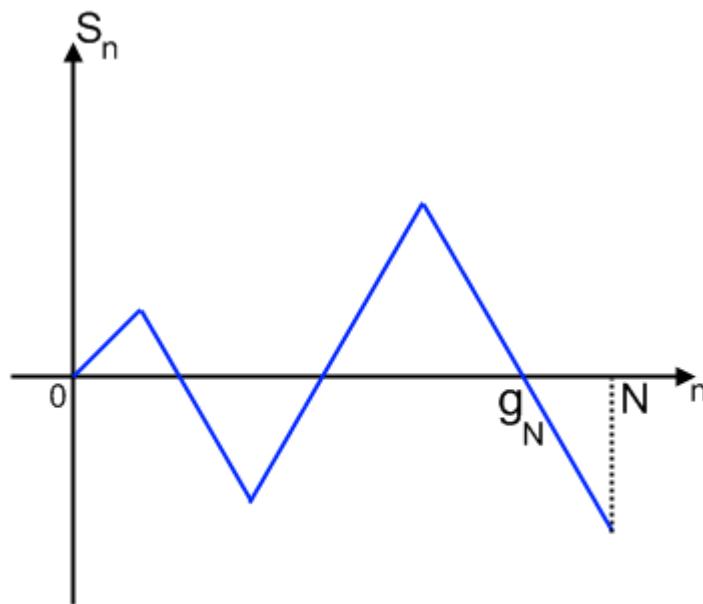
Integrating on  $z$ , we receive

$$F_{\tau-a} = \int_0^{+\infty} I(F_{\tau-a} \geq z) dz = - \sum_{k=2}^{\tau-a} H(F_{k-1}) \Delta S_k.$$

□

**3. Case**  $F_{g_N} = \max_{0 \leq k \leq g_N} S_k,$

$$g_N = \sup\{0 < k \leq N : S_k = 0\}.$$



**Theorem 3.** For  $F_{g_N}$  we have the stochastic representation

$$\max_{0 \leq k \leq g_N} S_k = \sum_{k=1}^{\left[\frac{N}{2}\right]} P(F_N \geq 2k) + \sum_{k=1}^N L_{N-k+1}(F_{k-1}, S_{k-1}) \Delta S_k - \sum_{k=1}^N M_{N-k+1}(S_{k-1})(F_{k-1} - F_{g_{k-1}}) \Delta S_k,$$

where

$$L_{N-k+1}(F_{k-1}, S_{k-1}) = \sum_{n=F_{k-1}-S_{k-1}+1}^{N-k+1} \sum_{m=0}^{\left[\frac{N-k+1-n}{2}\right]} \sum_{l=0}^{N-k+1-2m} (-1)^m \binom{\frac{1}{2}}{m}$$

$$\times\,\varphi_l(2n+S_{k-1}),$$

$$M_{N-k+1}(S_{k-1}) = \sum_{l,m:\ l+m\leqslant \left[\frac{N-k+1}{2}\right]} (-1)^m \binom{\frac{1}{2}}{m} \varphi_{S_{k-1}+2l}(S_{k-1})$$

$$\times\,I(S_{k-1}\leqslant N-k+1-2(l+m)).$$

## Multiple representations

Denote

$$B^n = \{\bar{i} = (i_1 \dots i_n) \mid i_k \in \{-1, 1\}, k = 1 \dots n\},$$

$$U_n^m = \{(t_1, \dots, t_m) : 1 \leq t_1 < \dots < t_m \leq n, t_i \in \mathbb{N}\}.$$

**Definition 1.** Let  $h: \mathbb{R}^n \rightarrow \mathbb{R}$ .

**The difference quotient of order  $n$**

$$D^{(1)}h(x_1, \dots, x_n) = \frac{h(1, x_2, \dots, x_n) - h(-1, x_2, \dots, x_n)}{2},$$

$$D^{(k)}h(x_1, \dots, x_n) = \frac{1}{2} D^{(k-1)}(h(x_1, \dots, x_{k-1}, 1, x_{k+1}, \dots, x_n)$$

$$-h(x_1, \dots, x_{k-1}, -1, x_{k+1}, \dots, x_n)), k \leq n.$$

For instance,  $D^{(2)}h(x_1, x_2) = 1/4(h(1, 1) - h(-1, 1) - h(1, -1) + h(-1, -1))$ .

**Proposition 1.** *For any  $n \in \mathbb{N}$  and for any function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  we have*

$$\frac{1}{2^n} \sum_{\bar{i} \in B^n} i_{t_1} \dots i_{t_k} f(\bar{i}) = D^{(k)} \left( \mathbb{E} [f(\bar{\xi}) | \xi_{t_1} = x_1, \dots, \xi_{t_k} = x_k] \right),$$

where  $1 \leq t_1 < \dots < t_k \leq n$ ,  $\bar{\xi} = (\xi_1, \dots, \xi_n)$ .

**Proof** induction on  $k$ .

**Lemma 2.** Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}$ . Then we have

$$f(\bar{\xi}) = Ef(\bar{\xi}) + \sum_{k=1}^n \sum_{\bar{t} \in U_n^k} D^{(k)} \left( E \left[ f(\bar{\xi}) \mid \xi_{t_1} = x_1, \dots, \xi_{t_k} = x_k \right] \right) \Delta S_{t_1} \dots \Delta S_{t_k},$$

where  $\bar{\xi} = (\xi_1, \dots, \xi_n)$ .

**Corollary 1.** For  $F_N = \max_{0 \leq k \leq N} S_k$  we have the representation

$$F_N = EF_N + \sum_{k=1}^N \sum_{1 \leq t_1 < \dots < t_k \leq N} D^{(k)} \left( E \left[ F_N \mid \xi_{t_1} = x_1, \dots, \xi_{t_k} = x_k \right] \right) \Delta S_{t_1} \dots \Delta S_{t_k}.$$

## 1. Multiple representation of functional $F_N = \max_{0 \leq k \leq N} S_k$

**Theorem 4.** For  $F_N = \max_{0 \leq k \leq N} S_k$  we have the following multiple stochastic representation:

$$F_N = E F_N + \sum_{k=1}^N E \left[ G_{N-k+1} (F_{k-1} - S_{k-1}) \right] \Delta S_k \\ + \sum_{m=2}^N \sum_{1 \leq k_1 < \dots < k_m \leq N} c(k_1, \dots, k_m) \Delta S_{k_1} \dots \Delta S_{k_m},$$

where

$$c(k_1, \dots, k_m) = D^{(m-1)} (E \left[ G_{N-k_m+1} (F_{k_m-1} - S_{k_m-1}) | \xi_{k_1} = x_1, \dots, \xi_{k_{m-1}} = x_{m-1} \right]).$$

Proof.

Using the theorem 1, we have

$$F_N = \mathbb{E}F_N + \sum_{k=1}^N G_{N-k+1}(F_{k-1} - S_{k-1})\Delta S_k. \quad (6)$$

Apply lemma 3 for  $G_{N-k+1}(F_{k-1} - S_{k-1})$ :

$$G_{N-k+1}(F_{k-1} - S_{k-1}) = \mathbb{E}[G_{N-k+1}(F_{k-1} - S_{k-1})]$$

$$+ \sum_{l=2}^k \sum_{\bar{t} \in U_{k-1}^{l-1}} c(t_1, \dots, t_{l-1}, k) \Delta S_{t_1} \dots \Delta S_{t_{l-1}}.$$

Substituting the computed representation in (6), we get the statement of the theorem.

□

## 2. Multiple representation of functional $F_{\tau-a} = \max_{0 \leq k \leq \tau-a} S_k$

**Theorem 5.** For  $F_{\tau-a}$  we have the stochastic representation:

$$F_{\tau-a} = - \sum_{k=2}^{\tau-a} \mathbb{E}[H(F_{k-1})] \Delta S_k + - \sum_{m=2}^{\tau-a} \sum_{1 \leq k_1 < \dots < k_m} h(k_1, \dots, k_m) I(\tau-a \geq k_m) \Delta S_{k_1} \dots \Delta S_{k_m},$$

where

$$h(k_1, \dots, k_m) = D^{(m-1)} \left( \mathbb{E} \left[ H(F_{k_m-1}) \mid \xi_{k_1} = x_1, \dots, \xi_{k_{m-1}} = x_{m-1} \right] \right).$$

**Thank you for your attention**

# Appendix

## Single representations

1. Case  $F_N = \max_{0 \leq k \leq N} S_k$

**Theorem 6.** For  $F_N = \max_{0 \leq k \leq N} S_k$  we have the following stochastical representation:

$$\max_{0 \leq n \leq N} S_n = E(\max_{0 \leq n \leq N} S_n) + \sum_{k=1}^N G_{N-k+1}(F_{k-1} - S_{k-1}) \Delta S_k, \quad (7)$$

where

$$G_{N-k+1}(F_{k-1} - S_{k-1}) = \sum_{r=F_{k-1}-S_{k-1}+1}^{N-k+1} \sum_{l=0}^{\left[\frac{N-k+1-r}{2}\right]} (-1)^{\left[\frac{N-k+1-r}{2}\right]-l} \\ \times C_{-\frac{1}{2}}^{\left[\frac{N-k+1-r}{2}\right]-l} \varphi_{r+2l}(r),$$

$$\varphi_k(m) = \frac{m}{k2^k} C_k^{\frac{m+k}{2}}, \quad \mathbb{E}(\max_{0 \leq n \leq N} S_n) \sim \sqrt{\frac{2N}{\pi}} \quad \text{by} \quad N \rightarrow \infty.$$

Доказательство. Let  $\xi \geq 0$  is a random variable. Then

$$\xi = \int_0^{+\infty} I(\xi \geq a) da. \text{ Hence}$$

$$F_N = \int_0^{+\infty} I(\max_{0 \leq k \leq N} S_k \geq a) da = \int_0^{+\infty} I(N_a \leq N) da,$$

where  $N_a = \inf\{k \geq 0 : S_k \geq a\} = \inf\{k \geq 0 : S_k = \lceil a \rceil\} = N_{\lceil a \rceil}$ ,  $a \geq 0$ .

Consider the **discrete stochastical exponent** defines for  $\lambda \in (0, 1)$ :

$$\mathcal{E}_n(S) = \prod_{k=1}^n (1 + \lambda \Delta S_k) = (\sqrt{1 - \lambda^2})^n \left( \sqrt{\frac{1 + \lambda}{1 - \lambda}} \right)^{S_n}, \quad \mathcal{E}_0(S) := 1.$$

$\Delta \mathcal{E}_n(S) = \lambda \mathcal{E}_{n-1}(S) \Delta S_n$ , and  $\mathcal{E}_n(S)$  – martingale with respect to filtration  $(\mathcal{F}_n)_{n \geq 0}$ . From **the optional sampling theorem** (P-a. s.)

$$\mathbb{E}(\mathcal{E}_{N_a} | \mathcal{F}_n) = \mathcal{E}_{N_a \wedge n} = 1 + \sum_{k=1}^{N_a \wedge n} \Delta \mathcal{E}_k = 1 + \lambda \sum_{k=1}^{N_a \wedge n} \mathcal{E}_{k-1} \Delta S_k. \quad (8)$$

Because of  $S_{N_a} = \lceil a \rceil$ , then

$$\mathbb{E}(\mathcal{E}_{N_a} | \mathcal{F}_n) = \left( \sqrt{\frac{1 + \lambda}{1 - \lambda}} \right)^{\lceil a \rceil} \mathbb{E} \left[ (\sqrt{1 - \lambda^2})^{N_a} | \mathcal{F}_n \right]. \quad (9)$$

$\mathbb{E}\mathcal{E}_{N_a} = \mathbb{E}\mathcal{E}_n = 1$ , therefore

$$\mathbb{E}(\sqrt{1 - \lambda^2})^{N_a} = \left( \sqrt{\frac{1 + \lambda}{1 - \lambda}} \right)^{-\lceil a \rceil}. \quad (10)$$

From (8), (9), (10) we get

$$\begin{aligned} \mathbb{E}((\sqrt{1 - \lambda^2})^{N_a} | \mathcal{F}_n) &= \mathbb{E}(\sqrt{1 - \lambda^2})^{N_a} + \\ &+ \lambda \sum_{k=1}^{N_a \wedge n} \left( \sqrt{\frac{1 + \lambda}{1 - \lambda}} \right)^{-(\lceil a \rceil - S_{k-1})} (\sqrt{1 - \lambda^2})^{k-1} \Delta S_k. \end{aligned}$$

$\mathbb{P}(N_a = k) = \varphi_k(\lceil a \rceil)$ , so for all  $a \geq 0$

$$\sum_{l=0}^{+\infty} (\sqrt{1 - \lambda^2})^{\lceil a \rceil + 2l} \varphi_{\lceil a \rceil + 2l}(\lceil a \rceil) = \left( \sqrt{\frac{1 + \lambda}{1 - \lambda}} \right)^{-\lceil a \rceil}. \quad (11)$$

Denote  $\sqrt{1 - \lambda^2} =: s$

$$\Rightarrow \lambda = \sqrt{1 - s^2} = \sum_{m=0}^{+\infty} (-1)^m C_{\frac{1}{2}}^m s^{2m}. \quad (12)$$

From (11), (12) for all  $s \in (0, 1)$  we have

$$\begin{aligned} E(s^{N_a} | \mathcal{F}_n) &= E s^{N_a} + \sum_{k=1}^{N_a \wedge n} \sum_{l,m=0}^{+\infty} (-1)^m C_{\frac{1}{2}}^m \varphi_{\lceil a \rceil - S_{k-1} + 2l} (\lceil a \rceil - S_{k-1}) \\ &\quad \times s^{\lceil a \rceil - S_{k-1} + 2(m+l) + k - 1} \Delta S_k \end{aligned} \quad (13)$$

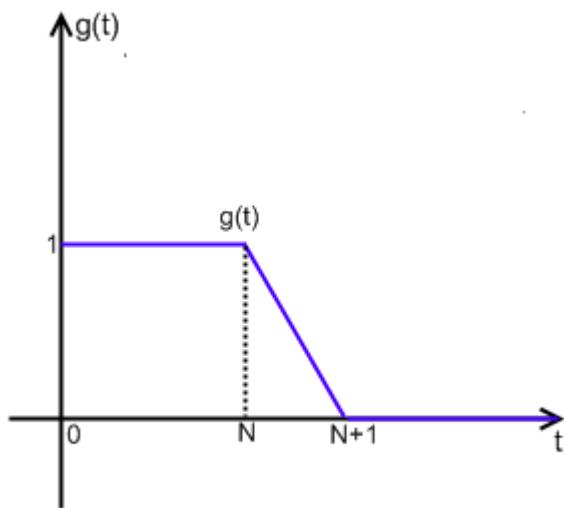
$$\left. \begin{array}{l} f_s(x) = s^x, x \geq 0, s \in (0, 1) \\ s = e^{-n}, n = 1, 2, \dots, \end{array} \right\} \Rightarrow (13) \text{ is valid for } \mathcal{L} = \{e^{-nx}\}_{n \geq 0}.$$

$$C_0[0, +\infty] = \{f \in C[0, +\infty] : f(x) \rightarrow 0 \text{ при } x \rightarrow +\infty\}$$

$\Downarrow_{(*)}$

$$\mathcal{L} = \{e^{-nx}\}_{n \geq 0} \text{ is complete in } C_0[0, +\infty].$$

Consider the function



(\*) Sedletski A. M. *Classes of analytic Fourier transformations and exponential approximations.*  
Moscow: Phismatlit, 2005, 504 p.

$$\left. \begin{array}{l} g(t) \in C_0[0, +\infty) \\ g(n) = I(n \leq N) \end{array} \right\} \Rightarrow (13) \text{ is valid for } I(n \leq N).$$

**Lemma 3.** For  $I(N_a \leq N)$  we have the representation (P-a.s.)

$$I(N_a \leq N) = P(N_a \leq N) + \sum_{k=1}^{N_a \wedge N} \sum_{l,m: l+m \leq \left[\frac{N-k+1}{2}\right]} (-1)^m C_{\frac{1}{2}}^m$$

$$\times \varphi_{\lceil a \rceil - S_{k-1} + 2l}(\lceil a \rceil - S_{k-1}) I(\lceil a \rceil - S_{k-1} \leq N - k + 1 - 2(l + m)) \Delta S_k.$$

$$F_N = \int_0^{+\infty} I(N_a \leq N) da = \int_0^{+\infty} \mathbb{P}(F_N \geq a) da +$$

$$+ \sum_{k=1}^N \sum_{l+m \leq \left[\frac{N-k+1}{2}\right]} \int_0^{S_{k-1}+N-k+1-2(l+m)} (-1)^m C_{\frac{1}{2}}^m \varphi_{\lceil a \rceil - S_{k-1} + 2l}(\lceil a \rceil - S_{k-1}) \\ \times I(k \leq N_a) da \Delta S_k =$$

$$\mathbb{E} F_N + \sum_{k=1}^N \sum_{l+m \leq \left[\frac{N-k+1}{2}\right]} (-1)^m C_{\frac{1}{2}}^m \left( \int_{F_{k-1} - S_{k-1}}^{N-k+1-2(l+m)} \varphi_{\lceil a \rceil + 2l}(\lceil a \rceil) da \right) \Delta S_k.$$

$$F_N = \mathbb{E}F_N + \sum_{k=1}^N \left( \sum_{l+m \leq \left[\frac{N-k+1}{2}\right]} \sum_{r=F_{k-1}-S_{k-1}+1}^{N-k+1-2(l+m)} (-1)^m C_{\frac{1}{2}}^m \varphi_{r+2l}(r) \right) \Delta S_k$$

$$= \mathbb{E}F_N + \sum_{k=1}^N \sum_{r=F_{k-1}-S_{k-1}+1}^{N-k+1} \left( \sum_{l+m \leq \left[\frac{N-k+1-r}{2}\right]} (-1)^m C_{\frac{1}{2}}^m \varphi_{r+2l}(r) \right) \Delta S_k.$$

For all  $x \in \mathbb{R}$

$$(-1)^n C_{x-1}^n = \sum_{m=0}^n (-1)^m C_x^m$$

$\Rightarrow$  theorem 6 is proved.

**Lemma 4.** For  $G_{N-k+1}(F_{k-1} - S_{k-1})$  we have the representation (P-a.s.)

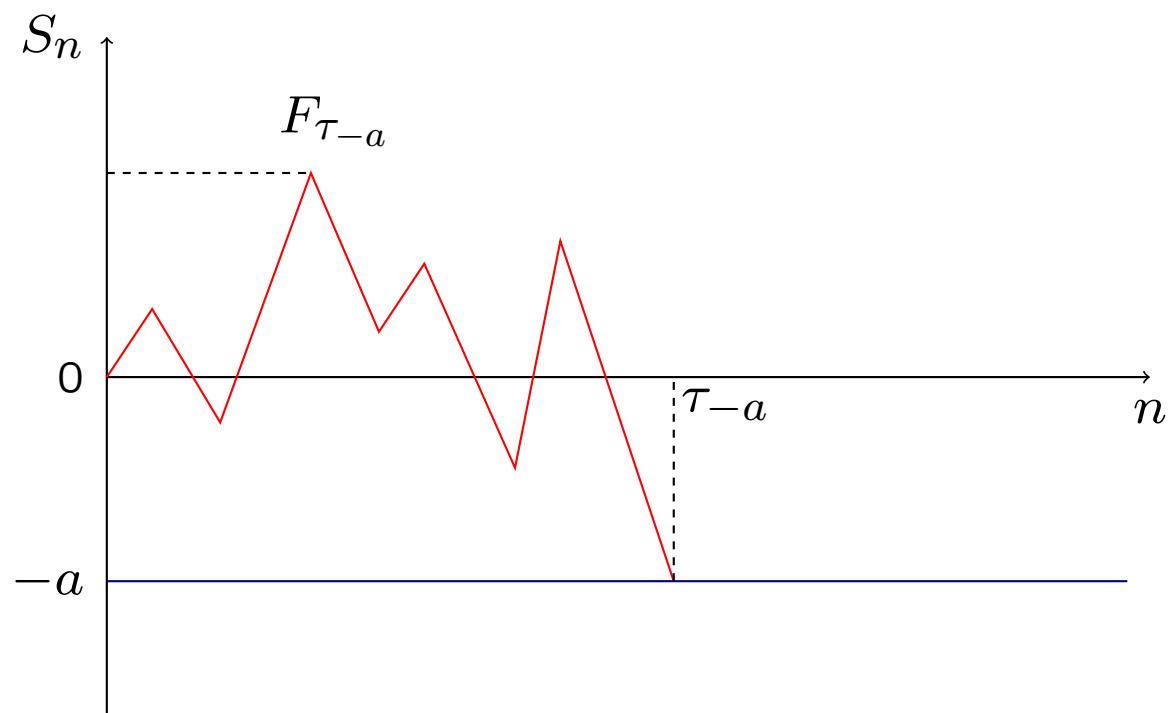
$$G_{N-k+1}(F_{k-1} - S_{k-1}) = \int_{F_{k-1}-S_{k-1}}^{N-k+1} (\mathbb{P}(F_{N-k+1} \geq a) - \mathbb{P}(F_{N-k+1-2\eta} \geq a)) da,$$

where  $\eta$  is a random variable, independent from  $\mathcal{F}_N$ , with distribution

$$\mathbb{P}(\eta=m) = (-1)^{m-1} C_{\frac{1}{2}}^m, \quad m = 1, 2 \dots$$

2. Case  $F_{\tau-a} = \max_{0 \leq k \leq \tau-a} S_k$ ,

$$\tau-a = \inf\{k \geq 0 : S_k = -a\}.$$



**Theorem 7.** For  $F_{\tau-a}$  we have the stochastical representation

$$\max_{0 \leq k \leq \tau-a} S_k = - \sum_{k=2}^{\tau-a} H(F_{k-1}) \Delta S_k,$$

where  $H(n) = \sum_{l=1}^n \frac{1}{a+l}$ ,  $F_k = \max_{0 \leq l \leq k} S_l$ .

Proof.

$$I(F_{\tau-a} \geq z) = I(N_z \leq \tau-a), \quad z > 0.$$



$$(a + \lceil z \rceil) I(F_{\tau-a} \geq z) = a + S_{\tau-a \wedge N_z}, \quad (14)$$

↓

$$S_{\tau-a \wedge N_z} + a = \sum_{k=1}^{\tau-a \wedge N_z} \Delta S_k + a = \sum_{k=1}^{\tau-a} I(k \leq N_z) \Delta S_k + a$$

$$= \sum_{k=1}^{\tau-a} I(F_{k-1} < z) \Delta S_k + a = - \sum_{k=2}^{\tau-a} I(F_{k-1} \geq z) \Delta S_k.$$

Therefore, from (14) we have

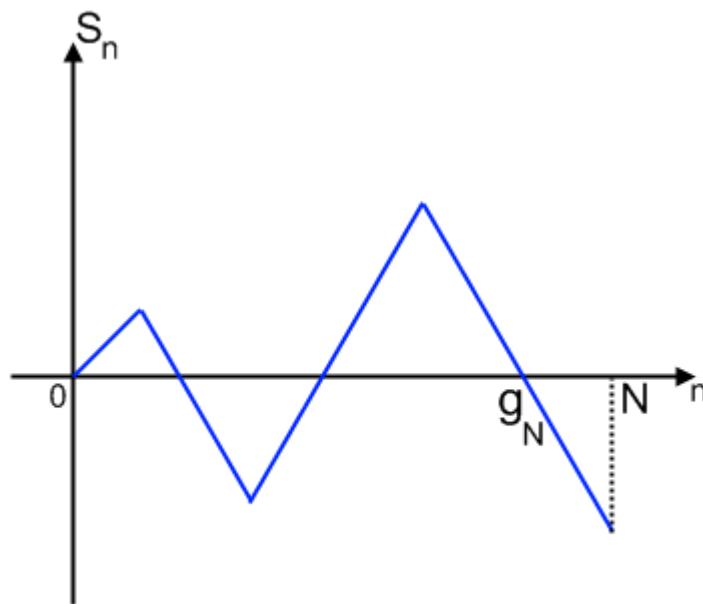
$$I(F_{\tau-a} \geq z) = - \sum_{k=2}^{\tau-a} \frac{I(F_{k-1} \geq z)}{a + \lceil z \rceil} \Delta S_k.$$

$$\begin{aligned} F_{\tau-a} &= \int_0^{+\infty} I(F_{\tau-a} \geq z) dz = - \int_0^{+\infty} \sum_{k=2}^{\tau-a} \frac{I(F_{k-1} \geq z)}{a + \lceil z \rceil} \Delta S_k dz \\ &= - \sum_{k=2}^{\tau-a} \left( \int_0^{F_{k-1}} \frac{dz}{a + \lceil z \rceil} \right) \Delta S_k = - \sum_{k=2}^{\tau-a} H(F_{k-1}) \Delta S_k. \end{aligned}$$

□

**3. Case**  $F_{g_N} = \max_{0 \leq k \leq g_N} S_k,$

$$g_N = \sup\{0 < k \leq N : S_k = 0\}.$$



**Theorem 8.** For  $F_{g_N}$  we have the stochastical representation

$$\max_{0 \leq k \leq g_N} S_k = \sum_{k=1}^{\left[\frac{N}{2}\right]} P(F_N \geq 2k) + \sum_{k=1}^N L_{N-k+1}(F_{k-1}, S_{k-1}) \Delta S_k$$

$$- \sum_{k=1}^N M_{N-k+1}(S_{k-1})(F_{k-1} - F_{g_{k-1}}) \Delta S_k,$$

where

$$L_{N-k+1}(F_{k-1},S_{k-1}) = \sum_{n=F_{k-1}-S_{k-1}+1}^{N-k+1} \sum_{m=0}^{\left[\frac{N-k+1-n}{2}\right]} \sum_{l=0}^{N-k+1-2m} (-1)^m C_{1/2}^m$$

$$\times\,\varphi_l(2n+S_{k-1}),$$

$$M_{N-k+1}(S_{k-1}) = \sum_{l,m:\; l+m\leqslant \left[\frac{N-k+1}{2}\right]} (-1)^m C_{\frac{1}{2}}^m \varphi_{S_{k-1}+2l}(S_{k-1})$$

$$\times\,I(S_{k-1}\leqslant N-k+1-2(l+m)).$$

Proof.

$$\begin{aligned} \max_{0 \leq n \leq g_N} S_n &= \int_0^{+\infty} I(\max_{0 \leq n \leq g_N} S_n \geq a) da \\ &= \int_0^{+\infty} I(N_a \leq g_N) da = \int_0^{+\infty} I(d_{N_a} \leq N) da, \end{aligned}$$

where

$$d_{N_a} = \inf\{k \geq N_a : S_k = 0\}.$$

$$d_{N_a} = N_a + \inf\{k \geq 0 : S_{k+N_a} = 0\} = N_a + \widehat{N}_{-a},$$

where

$$\widehat{N}_{-a} = \inf\{k \geq 0 : \widehat{S}_k = -\lceil a \rceil\},$$

$(\widehat{S}_k)_{k \geq 0}$  c  $\widehat{S}_k = S_{N_a+k} - \lceil a \rceil$  – random walk that independent from  $\mathcal{F}_{N_a}$ .

$$d_{N_a} = N_a + \widehat{N}_{-a}, \quad I(d_{N_a} \leq N) = I(\widehat{N}_{-a} \leq N - N_a).$$

$$\widehat{N}_{-a} \text{ и } N_a \text{ н. о. п.} \stackrel{\text{lemma 4}}{\Rightarrow} I(\widehat{N}_a(-\widehat{S}) \leq b) = I(\widehat{N}_{-a} \leq b);$$

Denote  $b = N - N_a$ .

$$I(d_{N_a} \leq N) = \sum_{l=0}^{+\infty} I(l \leq N - N_a) \varphi_l(\lceil a \rceil) - \sum_{k=(N_a+1)}^{d_{N_a} \wedge N} \sum_{l+m \leq \left[ \frac{N-k+1}{2} \right]} (-1)^m$$

$$\times C_{\frac{1}{2}}^m \varphi_{S_{k-1} + 2l}(S_{k-1}) I(S_{k-1} \leq N - k + 1 - 2(l+m)) \Delta S_k.$$

Using the relations below

$$\sum_{l=0}^N \mathbb{P}(N_a \leq N - l) \varphi_l(\lceil a \rceil) = \mathbb{P}(N_{2\lceil a \rceil} \leq N),$$

$$\begin{aligned} & \left[ \frac{N-k+1-n}{2} \right] - m \\ & \sum_{l=0}^{\left[ \frac{N-k+1-n}{2} \right] - m} \varphi_{n+2l}(n) \mathbb{P}(N_{n+x} \leq N - k + 1 - 2m - n - 2l) |_{x=S_{k-1}} = \\ & = \sum_{l=0}^{N-k+1-2m} \varphi_l(2n + S_{k-1}), \end{aligned}$$

we receive the statement of the theorem.

## Multiple representations

Denote

$$B^n = \{\bar{i} = (i_1 \dots i_n) \mid i_k \in \{-1, 1\}, k = 1 \dots n\},$$

$$U_n^m = \{(t_1, \dots, t_m) : 1 \leq t_1 < \dots < t_m \leq n, t_i \in \mathbb{N}\}.$$

**Definition 2.** Let  $h: \mathbb{R}^n \rightarrow \mathbb{R}$ .

**Difference quotient of order  $n$**

$$D^{(1)}h(x_1, \dots, x_n) = \frac{h(1, x_2, \dots, x_n) - h(-1, x_2, \dots, x_n)}{2},$$

$$D^{(k)}h(x_1, \dots, x_n) = \frac{1}{2} D^{(k-1)}(h(x_1, \dots, x_{k-1}, 1, x_{k+1}, \dots, x_n)$$

$$-h(x_1, \dots, x_{k-1}, -1, x_{k+1}, \dots, x_n)), k \leq n.$$

**Proposition 2.** For all  $n \in \mathbb{N}$  and for any function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$  we have

$$\frac{1}{2^n} \sum_{\bar{i} \in B^n} i_{t_1} \dots i_{t_k} f(\bar{i}) = D^{(k)} \left( E \left[ f(\bar{\xi}) \mid \xi_{t_1} = x_1, \dots, \xi_{t_k} = x_k \right] \right),$$

where  $1 \leq t_1 < \dots < t_k \leq n$ ,  $\bar{\xi} = (\xi_1, \dots, \xi_n)$ .

**Proof.** Induction on  $k$ .

When  $\mathbf{k=1}$  we have

$$\begin{aligned}
& \frac{1}{2^n} \sum_{\bar{i} \in B^n} i_{t_1} f(\bar{i}) = \frac{1}{2^n} \sum_{\bar{i} \in B^n: i_{t_1}=1} f(\bar{i}) - \frac{1}{2^n} \sum_{\bar{i} \in B^n: i_{t_1}=-1} f(\bar{i}) = \\
& = \frac{1}{2} \sum_{\bar{i} \in B^n: i_{t_1}=1} f(\bar{i}) P(\xi_1 = i_1, \dots, \widehat{\xi_{t_1}} \dots, \xi_n = i_n) - \frac{1}{2} \sum_{\bar{i} \in B^n: i_{t_1}=-1} f(\bar{i}) \\
& \times P(\xi_1 = i_1, \dots, \widehat{\xi_{t_1}} \dots, \xi_n = i_n) = \frac{1}{2} \left( E \left[ f(\bar{\xi}) |_{\xi_{t_1}=1} \right] - E \left[ f(\bar{\xi}) |_{\xi_{t_1}=-1} \right] \right) \\
& = D^{(1)} \left( E \left[ f(\bar{\xi}) |_{\xi_{t_1}=x} \right] \right).
\end{aligned}$$

So,

$$\begin{aligned}
& \frac{1}{2^n} \sum_{\bar{i} \in B^n} i_{t_1} \dots i_{t_{k+1}} f(\bar{i}) = \frac{1}{2} \left( D^{(k)} \left( \mathsf{E} \left[ f(\bar{\xi}) \mid \xi_{t_1} = x_1, \dots, \xi_{t_k} = x_k, \xi_{t_{k+1}} = 1 \right] \right) \right. \\
& \quad \left. - D^{(k)} \left( \mathsf{E} \left[ f(\bar{\xi}) \mid \xi_{t_1} = x_1, \dots, \xi_{t_k} = x_k, \xi_{t_{k+1}} = -1 \right] \right) \right) = \\
& \quad = D^{(k+1)} \left( \mathsf{E} \left[ f(\bar{\xi}) \mid \xi_{t_1} = x_1, \dots, \xi_{t_{k+1}} = x_{k+1} \right] \right).
\end{aligned}$$

□

**Lemma 5.** Let  $f: \mathbb{R}^n \rightarrow \mathbb{R}$ . Then we have the stochastical representation

$$f(\bar{\xi}) = \mathbb{E}f(\bar{\xi}) + \sum_{k=1}^n \sum_{\bar{t} \in U_n^k} D^{(k)} \left( \mathbb{E} [f(\bar{\xi}) | \xi_{t_1} = x_1, \dots, \xi_{t_k} = x_k] \right) \Delta S_{t_1} \dots \Delta S_{t_k},$$

where  $\bar{\xi} = (\xi_1, \dots, \xi_n)$ .

**Corollary 2.** For  $F_N = \max_{0 \leq k \leq N} S_k$  we have the stochastical representation

$$F_N = \mathbb{E}F_N + \sum_{k=1}^N \sum_{1 \leq t_1 < \dots < t_k \leq N} D^{(k)} \left( \mathbb{E} [F_N | \xi_{t_1} = x_1, \dots, \xi_{t_k} = x_k] \right) \Delta S_{t_1} \dots \Delta S_{t_k}.$$

## 1. Multiple representation of functional $F_N = \max_{0 \leq k \leq N} S_k$

**Theorem 9.** For  $F_N = \max_{0 \leq k \leq N} S_k$  we have the following multiple stochastic representation:

$$F_N = \mathbb{E}F_N + \sum_{k=1}^N \mathbb{E} [G_{N-k+1}(F_{k-1} - S_{k-1})] \Delta S_k$$

$$+ \sum_{m=2}^N \sum_{1 \leq k_1 < \dots < k_m \leq N} c(k_1, \dots, k_m) \Delta S_{k_1} \dots \Delta S_{k_m},$$

where

$$c(k_1, \dots, k_m) = D^{(m-1)} (\mathbb{E} [G_{N-k_m+1}(F_{k_m-1} - S_{k_m-1}) | \xi_{k_1} = x_1, \dots, \xi_{k_{m-1}} = x_{m-1}]).$$

Proof.

Apply lemma 6 for  $G_{N-k+1}(F_{k-1} - S_{k-1})$ .

$$G_{N-k+1}(F_{k-1} - S_{k-1}) = \mathbb{E}[G_{N-k+1}(F_{k-1} - S_{k-1})]$$

$$+ \sum_{m=1}^{k-1} \sum_{\bar{t} \in U_{k-1}^m} D^{(m)} \mathbb{E}[G_{N-k+1}(F_{k-1} - S_{k-1}) | \xi_{t_1} = x_1, \dots, \xi_{t_m} = x_m] \Delta S_{t_1} \dots \Delta S_{t_m}.$$

$$m = l - 1$$

$$G_{N-k+1}(F_{k-1} - S_{k-1}) = \mathbb{E}[G_{N-k+1}(F_{k-1} - S_{k-1})] \tag{15}$$

$$+ \sum_{l=2}^k \sum_{\bar{t} \in U_{k-1}^{l-1}} c(t_1, \dots, t_{l-1}, k) \Delta S_{t_1} \dots \Delta S_{t_{l-1}}.$$

Substituting (15) in formula (7), we have

$$F_N = \mathbb{E}F_N + \sum_{k=1}^N G_{N-k+1}(F_{k-1} - S_{k-1})\Delta S_k = \mathbb{E}F_N + \sum_{k=1}^N \mathbb{E}[G_{N-k+1}(F_{k-1} - S_{k-1})]\Delta S_k + \sum_{k=1}^N \sum_{l=2}^k \sum_{\bar{t} \in U_{k-1}^{l-1}} c(t_1, \dots, t_{l-1}, k)\Delta S_{t_1} \dots \Delta S_{t_{l-1}} \Delta S_k =$$

$$= \mathsf{E}F_N + \sum_{k=1}^N \mathsf{E}[G_{N-k+1}(F_{k-1} - S_{k-1})]\Delta S_k + \sum_{l=2}^N \sum_{k=l}^N \sum_{\bar{t} \in U_{k-1}^{l-1}} c(t_1,$$

$$\dots, t_{l-1}, k) \Delta S_{t_1} \dots \Delta S_{t_{l-1}} \Delta S_k = \mathsf{E}F_N + \sum_{k=1}^N \mathsf{E}[G_{N-k+1}(F_{k-1} - S_{k-1})]\Delta S_k$$

$$+ \sum_{l=2}^N \sum_{\bar{t} \in U_N^l} c(t_1, \dots, t_l) \Delta S_{t_1} \dots \Delta S_{t_l}.$$

□

## 2. Multiple representation of functional $F_{\tau-a} = \max_{0 \leq k \leq \tau-a} S_k$

**Theorem 10.** For  $F_{\tau-a}$  we have the multiple stochastic representation

$$F_{\tau-a} = - \sum_{k=2}^{\tau-a} \mathbb{E}[H(F_{k-1})] \Delta S_k$$

$$- \sum_{m=2}^{\tau-a} \sum_{1 \leq k_1 < \dots < k_m} h(k_1, \dots, k_m) I(\tau-a \geq k_m) \Delta S_{k_1} \dots \Delta S_{k_m},$$

where

$$h(k_1, \dots, k_m) = D^{(m-1)} \left( \mathbb{E} \left[ H(F_{k_{m-1}}) \mid \xi_{k_1} = x_1, \dots, \xi_{k_{m-1}} = x_{m-1} \right] \right).$$

Proof.

From lemma 6 we have

$$\begin{aligned} H(F_{k-1}) &= \mathbb{E}[H(F_{k-1})] + \\ &+ \sum_{m=1}^{k-1} \sum_{\bar{t} \in U_{k-1}^m} D^{(m)} \left( \mathbb{E} \left[ H(F_{k-1}) \mid \xi_{t_1} = x_1, \dots, \xi_{t_m} = x_m \right] \right) \Delta S_{t_1} \dots \Delta S_{t_m} = \\ &= \mathbb{E}[H(F_{k-1})] + \sum_{l=2}^k \sum_{\bar{t} \in U_{k-1}^{l-1}} h(t_1, \dots, t_{l-1}, k) \Delta S_{t_1} \dots \Delta S_{t_{l-1}}. \end{aligned}$$

Using theorem 7, we get the statement of the initial theorem.