

1st of August, 1993.

Some remarks on random scaling (: 1^{er} essai)

1)

In this Note, we are interested in the following question:

let $(V(t), t \geq 0)$ be a process (valued possibly in \mathbb{R}^n , or in an infinite dimensional vector space E), which ~~is~~ has the scaling property: for ~~any~~ every $c > 0$, $(V(ct), t \geq 0) \stackrel{\text{(law)}}{=} (cV(t), t \geq 0)$.

In particular, the distribution of $\frac{1}{t} V(t)$ does not depend on t ; the question we are interested in is: under which condition on a random variable h , taking values in $\mathbb{R}_+ \setminus \{0\}$, does one have:

$$(1) \quad \frac{1}{t} V(t) \stackrel{\text{(law)}}{=} \frac{1}{h} V(h).$$

Our motivation ~~to~~ to study this question comes from the following:

Example 1:

Take $V(t) \equiv A^+(t) \equiv \int_0^t ds \mathbb{1}_{\{B_s > 0\}}$,

and $h \equiv \tau(u) \equiv \inf \{ s : t_s > u \}$. Then:

$$(2) \quad \frac{1}{t} A^+(t) \stackrel{\text{(law)}}{=} \frac{1}{\tau(u)} A^+(\tau(u)), \quad \text{for every } t, u > 0.$$

In fact, instead of taking simply $V(t) = V_1(t) \equiv A^+(t)$, we can take:

$$V(t) = V_2(t) \equiv (A^+(t), t^2)^{(*)}$$

and we also have:

$$(3) \quad \frac{1}{t} V_2(t) \stackrel{\text{(law)}}{=} \frac{1}{\tau(u)} \left\{ V_2(\tau(u)) \right\}, \quad \text{for every } t, u > 0.$$

In this particular example, we would like to be able to replace $\tau(u)$ by a large number of random variables h .

(*) I preferred to restrict myself first to the 2-dimensional process, and a careful inspection of the proofs will show us how the results may be extended to our "full" process $(V(t), t \geq 0)$.

Throughout our discussion, we shall keep the following hypothesis:

- Hypothesis 3 (H₃) (i) $(H_t, t \geq 0)$ is a continuous increasing process, such that, for every t , H_t is $V(t)$ measurable;
- (ii) the pair (H, V) enjoys the scaling property:

for every $c > 0$, $(H_{ct}; V(ct); t \geq 0) \stackrel{\text{(law)}}{=} c(H_t; V(t); t \geq 0)$.

We now have the following:

Proposition 1: For every $f: \mathbb{R}_+ \times E \rightarrow \mathbb{R}_+$, we have:

$$(4) \quad E \left[\frac{dH_t}{H_t} f(H_t, V(t)) \right] = \frac{dt}{t} E \left[f\left(\frac{t}{h_1}, t \frac{V(h_1)}{h_1}\right) \right],$$

where $h_u \equiv \inf \{ t : H_t \geq u \}$.

Proof: Consider $\varphi: \mathbb{R}_+ \times \mathbb{R}_+ \times E \rightarrow \mathbb{R}_+$; then, we have:

$$\begin{aligned}
 E \left[\int_0^\infty \frac{dH_t}{H_t} \varphi(t, H_t, V(t)) \right] &= E \left[\int_0^\infty \frac{dx}{x} \varphi(h_x, x, V(h_x)) \right] \\
 &\quad \text{(by time-changing)} \\
 &= \int_0^\infty \frac{dx}{x} E \left[\varphi(x h_1, x, x V(h_1)) \right] \\
 &\quad \text{(by scaling)} \\
 &= \int_0^\infty \frac{dt}{t} E \left[\varphi\left(t, \frac{t}{h_1}, t \frac{V(h_1)}{h_1}\right) \right] \\
 &\quad \text{(taking } t = x h_1 \text{)}.
 \end{aligned}$$

Taking now $\varphi(t, \alpha, \beta) = g(t) f(\alpha, \beta)$, for a generic g , we obtain (4) \square

Corollary: Under our hypothesis ~~(H)~~^(H), the two following properties (5) and (6) are equivalent:

(5) $(H_1, V(1)) \stackrel{\text{(law)}}{=} \left(\frac{1}{h_1}, \frac{V(h_1)}{h_1} \right)$

(6) $E [dH_t | V(t)] = \frac{dt}{t} H_t.$

Proof: (6) \Rightarrow (5): As a consequence of (6) and (4), we get:

dt a.s., $E \left[f(t H_1, t V(1)) \right] = E \left[f\left(\frac{t}{h_1}, t \frac{V(h_1)}{h_1}\right) \right]$

from which we deduce (5).

(5) \Rightarrow (6): We want to show:

(7) $E \left[\int_0^\infty \frac{dH_t}{H_t} \varphi(t, V(t)) \right] = E \left[\int_0^\infty \frac{dt}{t} \varphi\left(t, \frac{V(h_1)}{h_1}\right) \right]$

for every $\varphi: \mathbb{R}_+ \times E \rightarrow \mathbb{R}_+$.

But, from (4), we know that the left-hand side of (7) is equal to:

$$\begin{aligned} E \left[\int_0^\infty \frac{dt}{t} \varphi\left(t, \frac{V(h_1)}{h_1}\right) \right] &= E \left[\int_0^\infty \frac{dt}{t} \varphi(t, t V(1)) \right], \text{ from (5)} \\ &= E \left[\int_0^\infty \frac{dt}{t} \varphi(t, V(t)) \right], \end{aligned}$$

using the scaling property of V. \square .

Notation:

If the properties (5) and/or (6) are satisfied, we will say that

H is V-admissible, or that V is h-stable.

The following proposition gives a recipe for creating "new" V -admissible processes ~~not~~ from "old ones".

Proposition 2: Assume that ~~the~~ n increasing processes H^1, H^2, \dots, H^n satisfy the hypothesis (H) with respect to V , and that moreover, each of them is V -admissible.

Assume moreover that:

for every c , $(H_{ct}^1, H_{ct}^2, \dots, H_{ct}^n; V(ct); t \geq 0)$

$$\stackrel{\text{(law)}}{=} (cH_t^1, cH_t^2, \dots, cH_t^n; cV(t); t \geq 0)$$

then, if $f: (\mathbb{R}_+)^n \rightarrow \mathbb{R}_+$ is a C^1 function, which is homogeneous of order (1), i.e.:

$$(8) \quad f(cx_1, \dots, cx_n) = c f(x_1, \dots, x_n),$$

then, the process:

$$H_t = f(H_t^1, H_t^2, \dots, H_t^n) \quad \text{is } V\text{-admissible.}$$

Proof: a) First, the hypothesis about $(H^1, H^2, \dots, H^n, V)$ made in this proposition and f

imply that (H, V) satisfies the hypothesis (H).

b) We now want to prove that H satisfies (6).

We have:

$$dH_t = \sum_{i=1}^n f'_i(H_t^1, H_t^2, \dots, H_t^n) dH_t^i$$

and so:
$$E[dH_t | V(t)] = \sum_{i=1}^n f'_i(H_t^1, H_t^2, \dots, H_t^n) E[dH_t^i | V(t)]$$

$$= \left(\frac{dt}{t}\right) \left(\sum_{i=1}^n f'_i(H_t^1, H_t^2, \dots, H_t^n) H_t^i\right)$$

since each of the H^i 's satisfies (6).

Now, from (8), we deduce that:

$$\sum_{i=1}^n f'_i(x_1, \dots, x_n) x_i = f(x_1, \dots, x_n),$$

hence, H satisfies (8) □

Before we consider some specific applications, let us prove some ~~more~~ ^{other} general facts.

Proposition 3: 1) Let $(H_t, t \geq 0)$ be a continuous increasing process such
that: for every $c > 0$, $(H_{ct}; t \geq 0) \stackrel{\text{(law)}}{=} (cH_t; t \geq 0)$
Then, H is H -admissible.

2) Let $(H_t, t \geq 0)$ and $(K_t, t \geq 0)$ be two continuous
increasing processes such that:
 for every $c > 0$, $(H_{ct}; K_{ct}; t \geq 0) \stackrel{\text{(law)}}{=} (c(H_t, K_t); t \geq 0)$
Define $V(t) = (H_t, K_t)$.

Then, H is V -admissible iff K is V -admissible.

3) Assume that $H_t = \int_0^t ds \theta(s)$, and that the process

(H, V) satisfies the hypothesis (H). Then, H is V -admissible iff:

$$(9) \quad \text{dt a.s.}, \quad E[\theta(t) | V(t)] = \frac{1}{t} H_t.$$

Proof: 1) If $V=H$, and H satisfies the scaling property, then,
 (5), which now reduces to:

$$H_1 \stackrel{\text{(law)}}{=} \frac{1}{h_1} \text{ is satisfied.}$$

2) In this case, since H is V -admissible, we have from (5):

$$(10) \quad (H_1, K_1) \stackrel{\text{(law)}}{=} \left(\frac{1}{h_1}; \frac{K h_1}{h_1} \right),$$

I have difficulties to prove this.

and we want to prove: (11) $(K_1, H_1) \stackrel{\text{(law)}}{=} \left(\frac{1}{k_1}, \frac{Hk_1}{k_1} \right)$.

Remark that (10), resp: (11), is equivalent to:

$$(10') \quad \left(H_1, \frac{K_1}{H_1} \right) \stackrel{\text{(law)}}{=} \left(\frac{1}{h_1}, K_{h_1} \right); \quad (11') \quad \left(K_1, \frac{H_1}{K_1} \right) \stackrel{\text{(law)}}{=} \left(\frac{1}{k_1}, H_{k_1} \right)$$

3) This is immediate as a consequence of the equivalence of (5) and (6)

Application: Let us consider again Example 1, with the 2-dimensional process $V \equiv V_2$.

a) The identity in law (3) tells us that l^2 is V_2 -admissible; hence, from Proposition 3, part 2), A^+ is also V_2 -admissible.

[Important note: Even if the proof of Prop. 3, part 2) cannot be completed, ~~it is~~ in particular, if this assertion is wrong!!, we can prove the result directly; what is less clear for me is that we can also replace V_2 by our original process

$$V(t) \equiv (V_1(t), V_2(t), \dots)$$

b) Define $A^-(t) \equiv t - A^+(t)$; it is a consequence of Proposition 2 that A^- is V_2 -admissible. (take: $H^1(t) \equiv t$, $H^0(t) \equiv A^+(t)$, and $f(x, y) = x - y$)

c) As an immediate consequence of the above discussion, we obtain ~~the above~~ the following examples (among many) of random times τ for which V_2 is τ -~~admissible~~ stable.

$$h = \inf \{ t : H_t > 1 \}, \quad \text{with:}$$

$$(i) H_t = aA_t^+ + bA_t^- + cL_t^2 \quad (a, b, c \geq 0); \quad (ii) H_t = (A_t^+ A_t^- L_t^2)^{1/3};$$

$$(iii) H_t = (aA_t^+ + bA_t^-)^{1/2} L_t \quad \dots$$

A negative example.

It may also be interesting to give some example of a 2-dimensional process $(V(t) \equiv (H(t), K(t)); t \geq 0)$ such that (H, V) satisfies (\mathcal{H}) , but H is not V -admissible.

This is the case with:

$$H_t = S_t^2; \quad K_t = L_t^2.$$

Indeed, if K were V -admissible, we would have:

$$(11?) \quad \frac{S_t^2}{\sigma(u)} \stackrel{\text{(law)}}{=} S_1^2, \quad z(u) \equiv \inf \{ t : L_t > u \}$$

or putting both sides upside down:

$$(12?) \quad \frac{\sigma(u)}{S_t^2} \stackrel{\text{(law)}}{=} \frac{1}{S_1^2} \left(\stackrel{\text{(law)}}{=} \sigma(1) \right).$$

[Well-known identity]

However, Knight's identity tells us precisely how wrong (12?) is!! Indeed, we have:

Another negative example :

$$V(t) \equiv (A^+(t), S_t^2)$$

If $H_t \equiv S_t^2$ were V -admissible, we would have:

$$(13?) \quad \frac{A^+(T_a)}{T_a} \stackrel{(\text{law})}{=} A^+(1) \quad (\text{which is a sine}).$$

or putting both sides upside down:

$$(14?) \quad \frac{T_a}{A^+(T_a)} \stackrel{(\text{law})}{=} \frac{1}{A^+(1)} \stackrel{(\text{law})}{=} (1+c^2).$$

A sufficient condition for H to be V -admissible

Suppose that : $V = (H, W)$, where W is independent of H , and (H, W) enjoys the scaling property. ($\Leftrightarrow (H, V)$ satisfies (\mathcal{H})).
Then, certainly:

$$(H_1, V(1)) \stackrel{(\text{law})}{=} \left(\frac{1}{h_1}, \frac{V(h_1)}{h_1} \right)$$

An interesting question seems to be: suppose that, under the general hypothesis

(\mathcal{H}) , H is V -admissible; then, does there exist a skew-product decomposition $V \equiv (H, W)$ of V with respect to H ??

On two-dimensional processes $(V(t) \equiv (H(t), K(t)); t \geq 0)$.

Assume that V satisfies the scaling property, and that H (and, therefore (?) K , by Proposition 3, 2)) is V -admissible.

Then, is it true that every process L such that (L, V) satisfies (\mathcal{H}) is V -admissible??

August 3rd, 1993.Some remarks on random scaling (: 2nd essai).

Below, I show that Proposition 1 of [July 29th], and Prop 1 of [Aug. 1st] can be assembled together in order to give ~~us~~ some better understanding of the laws ~~of~~ P^h of scaled Brownian motion:

$$\left(\frac{1}{\sqrt{h}} B_{sh}; s \leq 1 \right),$$

where: $h \equiv h_1 \equiv \inf \{u: H_u > 1\}$, and $(H_t, t \geq 0)$ is a process which scales jointly with Brownian motion; more precisely:

$$(1) \quad (H_{ct}, B_{ct}; t \geq 0) \stackrel{\text{law}}{=} (cH_t, \sqrt{c}B_t; t \geq 0).$$

First, from the identity (2.c.2), the distribution P^h , which is given by:

$$E^h [F(X_s; s \leq 1)] \stackrel{\text{def}}{=} E \left[F \left(\frac{1}{\sqrt{h}} B_{sh}; s \leq 1 \right) \right]$$

satisfies:

$$(2) \quad \left(\frac{dt}{t} \right) E^h [F(X_s, s \leq 1)] = E \left[\frac{dH_t}{H_t} F \left(\frac{1}{\sqrt{t}} B_{st}; s \leq 1 \right) \right]$$

Now, let us assume furthermore that, for every t , H_t is $\mathcal{V}(t)$ measurable, or, even better: \mathcal{G}_t measurable, and that:

$$(3) \quad E [dH_t | \mathcal{G}_t] = \frac{dt}{t} H_t.$$

Then, we deduce from (2) that:

$$(4) \quad P^h | \mathcal{G}_1 = P | \mathcal{G}_1.$$

We know that the hypothesis made on H is satisfied by:

$$H^+ = A^+, \quad H^- = A^-, \quad \text{and} \quad H^\# = l^2,$$

and I shall denote the corresponding probabilities P^h by:
 P^+ , P^- , and $P^\#$.

Now, let us consider:

$$(5) \quad H_t = f(A_t^+, A_t^-, l_t^2),$$

where f is an increasing continuous function in (x, y, z) , which satisfies:

$$f(cx, cy, cz) = c f(x, y, z).$$

Then, we have the following

Theorem:

If H is defined by (5), then:

$$(4) \quad P^h | e_1 = P | e_1 \quad \underline{\text{holds}}, \quad \underline{\text{and, furthermore:}}$$

$$(6) \quad P^h(\cdot | e_1) = \alpha P^+(\cdot | e_1) + \beta P^-(\cdot | e_1) + \gamma P^\#(\cdot | e_1)$$

where: $\alpha = \frac{f'_x(A_1^+, A_1^-, l_1^2) A_1^+}{f(A_1^+, A_1^-, l_1^2)}$, $\beta = \frac{f'_y(A_1^+, A_1^-, l_1^2) A_1^-}{f(A_1^+, A_1^-, l_1^2)}$, $\gamma = \frac{f'_z(A_1^+, A_1^-, l_1^2) l_1^2}{f(A_1^+, A_1^-, l_1^2)}$

(Sequel to: 2nd exam / Aug. 3rd).

Put slightly differently, we have:

$$(7) \quad \boxed{P^h = \alpha \cdot P^+ + \beta \cdot P^- + \gamma \cdot P^\#},$$

where, if δ is a ≥ 0 random variable, and Q a probability, $\delta \cdot Q$ indicates the measure: $\Gamma \rightarrow \int_{\Gamma} \delta dQ$.

Comments: 1) P^+ , P^- and $P^\#$ are carried by disjoint sets, namely: $(X_1 > 0)$, $(X_1 < 0)$, $X_1 = 0$.

2) As a particular case, we can write P , the Wiener measure in terms of P^+ and P^- :

$$(8) \quad \boxed{P = A_1^+ \cdot P^+ + A_1^- \cdot P^-}$$

Note that, although it is also true that:

$$(9) \quad P = A_1^+ \cdot P + A_1^- \cdot P,$$

obviously

~~it is not true that:~~
nonetheless we have:

$$A_1^+ \cdot P \neq A_1^+ \cdot P^+, \text{ and } A_1^- \cdot P \neq A_1^- \cdot P^-$$

To see (8) quickly, we should write, instead of (9):

$$P = 1_{(B_1 > 0)} \cdot P + 1_{(B_1 < 0)} \cdot P,$$

and then:

$$P(\cdot | \mathcal{G}_1) = P(B_1 > 0 | \mathcal{G}_1) \frac{P((B_1 > 0) \cap \cdot | \mathcal{G}_1)}{P(B_1 > 0 | \mathcal{G}_1)} + \left(\text{same with } < 0 \text{ instead of } > 0 \right)$$

and we know that:

$$P(B_1 \in \mathbb{R}^\pm | \mathcal{G}_1) = A_1^\pm, \text{ and } P^\pm(\cdot | \mathcal{G}_1) = \frac{P((B_1 \in \mathbb{R}^\pm) \cap \cdot | \mathcal{G}_1)}{A_1^\pm}$$

3) As a second particular case, we may assume the function f to be of the form:

$$f(x, y, z) \equiv \tilde{f}(x+y, z),$$

that is:

$$H_t \equiv f(A_t^+, A_t^-, L_t^z) = \tilde{f}(t, L_t^z).$$

Now, since f is homogeneous of degree 1, we have:

$$f(t, z) = t \varphi\left(\frac{z}{t}\right), \text{ for a certain function } \varphi,$$

and, therefore:

$$f'_t(t, z) = \varphi\left(\frac{z}{t}\right) - \left(\frac{z}{t}\right) \varphi'\left(\frac{z}{t}\right).$$

From (6), (7) and (8), we obtain:

$$(9) \quad P^h(\cdot | \mathcal{G}_1) = \left[1 - (L_1^z) \left(\frac{\varphi'}{\varphi}\right) (L_1^z)\right] P(\cdot | \mathcal{G}_1) + (L_1^z) \left(\frac{\varphi'}{\varphi}\right) (L_1^z) P^\#(\cdot | \mathcal{G}_1)$$

It may be a better idea to write:

$$f(t, z) = z \psi\left(\frac{t}{z}\right), \quad \text{so that: } f'_t(t, z) = \psi'\left(\frac{t}{z}\right).$$

Now, ~~the~~ formula (9) can be written in the form:

$$(9') \quad P^h(\cdot | \mathcal{G}_1) = \left(\frac{\psi'}{\psi}\right)\left(\frac{1}{L_1^z}\right) \cdot P(\cdot | \mathcal{G}_1) + \left[1 - \left(\frac{\psi'}{\psi}\right)\left(\frac{1}{L_1^z}\right)\right] \cdot P^\#(\cdot | \mathcal{G}_1)$$

It may be interesting to look for a function ψ such that $g \equiv \sup\{t < 1: X_t = 0\}$ has a given law: $\int \delta(t) dt$ may be under P^h (at least, $\int \delta(t) dt$ would be the absolutely continuous component of the law of g).

Now, since, under P , we have: $l_1^{(2)} \stackrel{(law)}{=} g \cdot (2T)$,

with g are sine distributed, and T exponential, we can write: for every $u: [0,1] \rightarrow \mathbb{R}_+$,

$$\int_0^1 dt \delta(t) \stackrel{u(t)}{=} u(t) = \int_0^1 \frac{dt u(t)}{\pi \sqrt{t(1-t)}} \int_0^\infty dx e^{-2} \frac{\psi'}{\psi} \left(\frac{1}{2tx} \right), \text{ so that:}$$

$$(*) \quad \boxed{\gamma(t) = \frac{1}{\pi \sqrt{t(1-t)}} \int_0^\infty dx e^{-x} \left(\frac{\psi'}{\psi} \right) \left(\frac{1}{2tx} \right)}$$

$$= \frac{1}{2\pi \sqrt{t^3(1-t)}} \int_0^\infty dy e^{-\left(\frac{y}{2t}\right)} \left(\frac{\psi'}{\psi} \right) \left(\frac{1}{y} \right).$$

$$t = 1/u \Rightarrow \frac{1}{u^2} \gamma\left(\frac{1}{u}\right) = \frac{1}{2\pi \sqrt{u-1}} \int_0^\infty dy e^{-\left(\frac{yu}{2}\right)} \left(\frac{\psi'}{\psi} \right) \left(\frac{1}{y} \right).$$

$$u = v+1 / \quad 2\pi \frac{\sqrt{v}}{(v+1)^2} \gamma\left(\frac{1}{v+1}\right) = \int_0^\infty dy \exp\left(-\frac{y}{2}(v+1)\right) \left(\frac{\psi'}{\psi} \right) \left(\frac{1}{y} \right).$$

(*) At least, starting from this formula, with a function ψ such that: $0 \leq \frac{\psi'}{\psi} \leq 1$, we obtain a number of distributions for g under P_h ;
 maybe look at some particular cases:

$$\left(\frac{\psi'}{\psi} \right) \left(\frac{2}{3} \right) = \frac{1}{1+3^\alpha}, \dots /$$

Some remarks on random scaling (: 3^e essai).

August 4th, 1993.

This is a succession of comments on the 1st essai [1st Aug.].

1. Comments on Proposition 1 and its Corollary.

Here, I remark that the identity (5) or (6) amounts to some "weak" Markov property for the triple:

$$(V(t), H_t, dH_t),$$

where: $V(t)$ plays the role of the entire past,
 H_t the present, (in particular, H_t is measurable with respect to $V(t)$)
 dH_t the future

The Corollary on p. 3 of [1st essai] may be presented as follows:

Proposition: Under the hypothesis (H), the property:

$$(6) \quad E[dH_t | V(t)] = \frac{dt}{t} H_t$$

is equivalent to: (6') $E[dH_t | V(t)] = E[dH_t | H_t]$.

Proof: Obviously, (6) implies (6'), and, ~~consequently~~ ^{conversely}, if (6') is satisfied, then the scaling property of the process $(H_t, t \geq 0)$ together with the general identity (4) implies:

$$E\left[\frac{dH_t}{H_t} f(H_t)\right] = \frac{dt}{t} E\left[f\left(\frac{t}{h_1}\right)\right] = \frac{dt}{t} E[f(H_t)]$$

so that, we always have: $E[dH_t | H_t] = \frac{dt}{t} H_t$.

Consequently, (6') implies (6) □

Particular cases:

a) Assume that (4) is satisfied, and that $H_t = \int_0^t ds \theta(s)$. Then, (6), or (6'), is satisfied iff:

(9') dt a.s., $E[\theta(t) | V(t)] = E[\theta(t) | H_t]$.

(and, again, we know a priori that: $E[\theta(t) | H_t] = \frac{1}{t} H_t$ is true under the scaling hypothesis for the process $(H_t, t \geq 0)$).

b) If, even more particularly, $\theta(t)$ is ~~an~~ indicator of a random set, which is the case in the arc sine study, then (9') is equivalent to:

(9'') dt a.s., the triple $(V(t), H_t, \theta_t)$ is Markovian.

Remark: The above characterization shows that the skew product idea on p. 8 of [1st essai] is somewhat naive, although the scaling property (of V) jointly with the Markov property (9'') may lead to some independence result;

see, e.g., Lamperti's representation of semistable Markov processes.

c) of course, we may also write (9'') in the following equivalent way: for every measurable f,

(9''') $E[f(V(t)) | H_t, \theta_t] = E[f(V(t)) | H_t]$.

a.e.

$V(t)$ and θ_t are independent conditionally on H_t .

This suggests the following questions:

Question 1: In the Brownian case, how difficult, or how simple is it to prove that:

$$P(B_t > 0 | V(t)) = P(B_t > 0 | A_t^+),$$

or, equivalently:

$$E[f(V(t)) | A_t^+, (B_t > 0)] = E[f(V(t)) | A_t^+]$$

Question 2:

(A simple answer to Question 1 should shed some light on analogous prob. for perturbed reflection BM).

The preceding identity suggests strongly that one studies the distribution of $(B_u; u \leq t)$ or, only: A_t^+ , so that:

given either the pair: $(A_t^+, (B_t > 0))$

(i) one may recover the above identity;

(ii) one may find some "maximal" σ -fields \mathcal{G}^{\max} such that:

\mathcal{G}^{\max} and $(B_t > 0)$ are independent conditionally on A_t^+ .

2. A closer look at the two "negative" examples.

$$(2.a) \quad \boxed{V(t) = (S_t^2; t_t^2)} \quad ; \quad (2.b) \quad \boxed{V(t) = (A^+(t), S_t^2)}$$

(2.b) Introducing the notation $T_a = \inf\{t: B_t = a\}$, I_{\neq} will now

show that: $(13')$ $\frac{A^+(T_1)}{T_1} \stackrel{\text{(law)}}{\neq} A^+(1)$,

or equivalently: $(14')$ $\frac{T_1}{A^+(T_1)} \stackrel{\text{(law)}}{\neq} \frac{1}{A^+(1)} \left(\stackrel{\text{(law)}}{=} (1+C^2) \right)$

where C is a standard Cauchy variable -

Indeed, we have:

$$\frac{T_1}{A^+(T_1)} = 1 + \frac{A^-(T_1)}{A^+(T_1)}$$

and, in order to prove (14'), I need to show:

$$(15) \quad \frac{A^-(T_1)}{A^+(T_1)} \stackrel{\text{(law)}}{\neq} C^2$$