

A sequence of Albin type continuous martingales with Brownian marginals and scaling

Catherine Donati-Martin, David Baker, Marc Yor

▶ To cite this version:

Catherine Donati-Martin, David Baker, Marc Yor. A sequence of Albin type continuous martingales with Brownian marginals and scaling. 2010. hal-00471169

HAL Id: hal-00471169 https://hal.science/hal-00471169

Preprint submitted on 7 Apr 2010

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

A sequence of Albin type continuous martingales with Brownian marginals and scaling

D. Baker¹, C. Donati-Martin¹, and M. Yor¹²

Dedicated to Lester Dubins (1921-2010) to whom the third author owes a lot.

Summary. Closely inspired by Albin's method which relies ultimately on the duplication formula for the Gamma function, we exploit Gauss' multiplication formula to construct a sequence of continuous martingales with Brownian marginals and scaling.

Key words: Martingales, Brownian marginals

1 Motivation and main results

(1.1) Knowing the law of a "real world" random phenomena, i.e. random process, $(X_t, t \ge 0)$ is often extremely difficult and in most instances, one avails only of the knowledge of the 1-dimensional marginals of $(X_t, t \ge 0)$. However, there may be many different processes with the same given 1-dimensional marginals.

In the present paper, we make explicit a sequence of continuous martingales $(M_m(t), t \ge 0)$ indexed by $m \in \mathbb{N}$ such that for each m,

i) $(M_m(t), t \ge 0)$ enjoys the Brownian scaling property: for any c > 0,

$$(M_m(c^2t), t \ge 0) \stackrel{(law)}{=} (cM_m(t), t \ge 0)$$

ii) $M_m(1)$ is standard Gaussian.

Note that, combining i) and ii), we get, for any t > 0

$$M_m(t) \stackrel{(law)}{=} B_t,$$

UPMC Univ Paris 06 and CNRS, UMR 7599, Laboratoire de Probabilités et Modèles Aléatoires, 4, place Jussieu, 75252 Paris cedex 05. E-mail: david.baker@etu.upmc.fr, catherine.donati@upmc.fr

 $^{^{2}\,}$ Institut Universitaire de France

where $(B_t, t \ge 0)$ is a Brownian motion, i.e. M_m admits the same 1-dimensional marginals as Brownian motion.

(1.2) Our main result is the following extension of Albin's construction [1] from m = 1 to any integer m.

Theorem 1. Let $m \in \mathbb{N}$. Then, there exists a continuous martingale $(M_m(t), t \ge 0)$ which enjoys i) and ii) and is defined as follows:

$$M_m(t) = X_t^{(1)} \dots X_t^{(m+1)} Z_m \tag{1}$$

where $(X_t^{(i)}, t \ge 0)$, for i = 1, ..., m+1, are independent copies of the solution of the SDE

$$dX_t = \frac{1}{m+1} \frac{dB_t}{X_t^m}; \quad X_0 = 0$$
 (2)

and, furthermore, Z_m is independent from $(X^{(1)}, \ldots, X^{(m+1)})$ and

$$Z_m \stackrel{(law)}{=} (m+1)^{1/2} \left(\prod_{j=0}^{m-1} \beta(\frac{1+2j}{2(m+1)}, \frac{m-j}{m+1}) \right)^{\frac{1}{2(m+1)}}$$
(3)

where $\beta(a,b)$ denotes a beta variable with parameter (a,b) with density

$$\frac{\Gamma(a+b)}{\Gamma(a)\Gamma(b)}x^{a-1}(1-x)^{b-1}1_{[0,1]}(x)$$

and the beta variables on the right-hand side of (3) are independent.

Remark: For m = 1, $Z_1 = \sqrt{2} \left(\beta(\frac{1}{4}, \frac{1}{2})\right)^{1/4}$ and we recover the distribution of $Y := Z_1$ given by (2) in [1].

(1.3) For the convenience of the reader, we also recall that, if one drops the continuity assumption when searching for martingales $(M(t); t \ge 0)$ satisfying i) and ii), then, the Madan-Yor construction [5] based on the "Azéma-Yor under scaling" method provides such a martingale.

Precisely, starting from a Brownian motion $(B_u, u \ge 0)$ and denoting $S_u = \sup_{s \le u} B_s$, introduce the family of stopping times

$$\tau_t = \inf\{u, S_u > \psi_t(B_u)\}$$

where ψ_t denotes the Hardy-Littlewood function associated with the centered Gaussian distribution μ_t with variance t, i.e.

$$\psi_t(x) = \frac{1}{\mu_t([x,\infty[)]} \int_x^\infty y \exp(-\frac{y^2}{2t}) \frac{dy}{\sqrt{2\pi t}}$$
$$= \sqrt{t} \exp(-\frac{x^2}{2t}) / \mathcal{N}(x/\sqrt{t})$$

where $\mathcal{N}(a) = \int_a^\infty \exp(-\frac{y^2}{2}) dy$. Then, $M_t = B_{\tau_t}$ is a martingale with Brownian marginals.

Another solution has been given by Hamza and Klebaner [4].

2 Proof of the theorem

Step 1: For $m \in \mathbb{R}$ and $c \in \mathbb{R}$, we consider the stochastic equation:

$$dX_t = c\frac{dB_t}{X_t^m}, \quad X_0 = 0.$$

This equation has a unique weak solution which can be defined as a timechanged Brownian motion

$$(X_t) \stackrel{(law)}{=} W(\alpha^{(-1)}(t))$$

where W is a Brownian motion starting from 0 and $\alpha^{(-1)}$ is the (continuous) inverse of the increasing process

$$\alpha(t) = \frac{1}{c^2} \int_0^t W_u^{2m} du.$$

We look for $k \in \mathbb{N}$ and c such that $(X_t^{2k}, t \ge 0)$ is a squared Bessel process of some dimension d. It turns out, by application of Itô's formula, that we need to take k=m+1 and $c=\frac{1}{m+1}$. Thus, we find that $(X_t^{2(m+1)}, t \ge 0)$ is a squared Bessel process with dimension $d=k(2k-1)c^2=\frac{2m+1}{m+1}$.

Note that the law of a BESQ(d) process at time 1 is well known to be that of $2\gamma_{d/2}$, where γ_a denotes a gamma variable with parameter a. Thus, we have:

$$|X_1| \stackrel{(law)}{=} \left(2\gamma_{\frac{2m+1}{2(m+1)}}\right)^{\frac{1}{2(m+1)}}$$
 (4)

Step 2: We now discuss the scaling property of the solution of (2). From the scaling property of Brownian motion, it is easily shown that , for any $\lambda > 0$, we get:

$$(X_{\lambda t}, t \ge 0) \stackrel{(law)}{=} (\lambda^{\alpha} X_t, t \ge 0)$$

with $\alpha = \frac{1}{2(m+1)}$, that is, the process $(X_t, t \ge 0)$ enjoys the scaling property of order $\frac{1}{2(m+1)}$.

Step 3: Consequently, if we multiply m+1 independent copies of the process $(X_t, t \ge 0)$ solution of (2), we get a process

$$Y_t = X_t^{(1)} \dots X_t^{(m+1)}$$

which is a martingale and has the scaling property of order $\frac{1}{2}$.

Step 4: Finally, it suffices to find a random variable Z_m independent of the processes $X_t^{(1)}, \ldots, X_t^{(m+1)}$ and which satisfies:

$$N \stackrel{(law)}{=} X_1^{(1)} \dots X_1^{(m+1)} Z_m \tag{5}$$

where N denotes a standard Gaussian variable. Note that the distribution of any of the $X_1^{(i)}$'s is symmetric. We shall take $Z_m \geq 0$; thus, the distribution of Z_m shall be determined by its Mellin transform $\mathcal{M}(s)$. From (5), $\mathcal{M}(s)$ satisfies:

$$\mathbb{E}[(2\gamma_{1/2})^{s/2}] = \left(\mathbb{E}[(2\gamma_{d/2})^{s/2(m+1)}]\right)^{m+1} \mathcal{M}(s)$$

with $d = \frac{2m+1}{m+1}$, that is:

$$2^{s/2} \frac{\Gamma(\frac{1+s}{2})}{\Gamma(\frac{1}{2})} = 2^{s/2} \left(\frac{\Gamma(\frac{d}{2} + \frac{s}{2(m+1)})}{\Gamma(\frac{d}{2})} \right)^{m+1} \mathcal{M}(s)$$

that is precisely:

$$\frac{\Gamma(\frac{1+s}{2})}{\Gamma(\frac{1}{2})} = \left(\frac{\Gamma(\frac{2m+1+s}{2(m+1)})}{\Gamma(\frac{2m+1}{2(m+1)})}\right)^{m+1} \mathcal{M}(s). \tag{6}$$

Now, we recall Gauss multiplication formula ([2], see also [3])

$$\Gamma(kz) = \frac{k^{kz-1/2}}{(2\pi)^{\frac{k-1}{2}}} \prod_{j=0}^{k-1} \Gamma(z + \frac{j}{k})$$
 (7)

which we apply with k = m + 1 and $z = \frac{1+s}{2(m+1)}$. We then obtain, from (7)

$$\frac{\Gamma(\frac{1+s}{2})}{\sqrt{\pi}} = \frac{(m+1)^{s/2}}{(2\pi)^{m/2}} \frac{1}{\sqrt{\pi}} \prod_{j=0}^{m} \Gamma(\frac{1+s+2j}{2(m+1)})$$
(8)

$$= (m+1)^{s/2} \prod_{j=0}^{m} \left(\frac{\Gamma(\frac{1+s+2j}{2(m+1)})}{\Gamma(\frac{1+2j}{2(m+1)})} \right)$$
(9)

since the two sides of (8) are equal to 1 for s = 0. We now plug (9) into (6) and obtain

$$(m+1)^{s/2} \prod_{j=0}^{m} \left(\frac{\Gamma(\frac{1+s+2j}{2(m+1)})}{\Gamma(\frac{1+2j}{2(m+1)})} \right) = \left(\frac{\Gamma(\frac{2m+1+s}{2(m+1)})}{\Gamma(\frac{2m+1}{2(m+1)})} \right)^{m+1} \mathcal{M}(s)$$
 (10)

We note that for j = m, the same term appears on both sides of (10), thus (10) may be written as:

$$(m+1)^{s/2} \prod_{j=0}^{m-1} \left(\frac{\Gamma(\frac{1+s+2j}{2(m+1)})}{\Gamma(\frac{1+2j}{2(m+1)})} \right) = \left(\frac{\Gamma(\frac{2m+1+s}{2(m+1)})}{\Gamma(\frac{2m+1}{2(m+1)})} \right)^m \mathcal{M}(s)$$
 (11)

In terms of independent gamma variables, the left-hand side of (11) equals:

$$(m+1)^{s/2}\mathbb{E}\left[\left(\prod_{j=0}^{m-1} \gamma_{\frac{1+2j}{2(m+1)}}^{(j)}\right)^{\frac{s}{2(m+1)}}\right]$$
(12)

whereas the right-hand side of (11) equals:

$$\mathbb{E}\left[\left(\prod_{j=0}^{m-1} \gamma_{\frac{1+2m}{2(m+1)}}^{(j)}\right)^{\frac{s}{2(m+1)}}\right] \mathcal{M}(s) \tag{13}$$

where the $\gamma_{a_j}^{(j)}$ denote independent gamma variables with respective parameters a_j .

Now, from the beta-gamma algebra, we deduce, for any $j \leq m-1$:

$$\gamma_{\frac{1+2j}{2(m+1)}}^{(j)} \overset{(law)}{=} \gamma_{\frac{1+2m}{2(m+1)}}^{(j)} \beta(\frac{1+2j}{2(m+1)}, \frac{m-j}{m+1}).$$

Thus, we obtain, again by comparing (12) and (13):

$$\mathcal{M}(s) = (m+1)^{s/2} \mathbb{E} \left[\left(\prod_{j=0}^{m-1} \beta(\frac{1+2j}{2(m+1)}, \frac{m-j}{m+1}) \right)^{\frac{s}{2(m+1)}} \right]$$

which entails:

$$\mathbb{E}[Z_m^s] = (m+1)^{s/2} \ \mathbb{E}\left[\left(\prod_{j=0}^{m-1} \beta(\frac{1+2j}{2(m+1)}, \frac{m-j}{m+1}) \right)^{\frac{s}{2(m+1)}} \right]$$

that is, equivalently,

$$Z_m \stackrel{(law)}{=} (m+1)^{1/2} \left(\prod_{j=0}^{m-1} \beta(\frac{1+2j}{2(m+1)}, \frac{m-j}{m+1}) \right)^{\frac{1}{2(m+1)}}$$

3 Some remarks about Theorem 1

3.1 A further extension

We tried to extend Theorem 1 by taking a product of independent martingales $X^{(i)}$, solution of (2) with different m_i 's. Here are the details of our attempt. We are looking for the existence of a variable Z such that the martingale

$$M(t) = \left(\prod_{j=0}^{p-1} X_t^{(m_j)}\right) Z$$

satisfies the properties i) and ii). Here $p, (m_j)_{0 \le j \le p-1}$ are integers and $X^{(m_j)}$ is the solution of the EDS (2) associated to m_j , the martingales being independent for j varying. In order that M enjoys the Brownian scaling property, we need the following relation

$$\sum_{j=0}^{p-1} \frac{1}{m_j + 1} = 1. \tag{14}$$

Following the previous computations, see (6), the Mellin transform $\mathcal{M}(s)$ of Z should satisfy

$$\frac{\Gamma(\frac{1+s}{2})}{\Gamma(\frac{1}{2})} = \left(\prod_{j=0}^{p-1} \frac{\Gamma(\frac{2m_j+1+s}{2(m_j+1)})}{\Gamma(\frac{2m_j+1}{2(m_j+1)})}\right) \mathcal{M}(s). \tag{15}$$

We recall (see (9)) the Gauss multiplication formula

$$\frac{\Gamma(\frac{1+s}{2})}{\sqrt{\pi}} = p^{s/2} \prod_{j=0}^{p-1} \left(\frac{\Gamma(\frac{1+s+2j}{2p})}{\Gamma(\frac{1+2j}{2p})} \right)$$
 (16)

To find $\mathcal{M}(s)$ from (15), (16), we give some probabilistic interpretation:

$$\frac{\Gamma(\frac{1+s+2j}{2p})}{\Gamma(\frac{1+2j}{2p})} = \mathbb{E}[\gamma_{(1+2j)/2p}^{s/2p}]$$

whereas

$$\frac{\Gamma(\frac{2m_j+1+s}{2(m_j+1)})}{\Gamma(\frac{2m_j+1}{2(m_i+1)})} = \mathbb{E}[\gamma_{(1+2m_j)/2(m_j+1)}^{s/2(m_j+1)}].$$

Thus, we would like to factorize

$$\gamma_{(1+2j)/2p}^{1/2p} \stackrel{(law)}{=} \gamma_{(1+2m_j)/2(m_j+1)}^{1/2(m_j+1)} z_{m_j,p}^{(j)}$$
(17)

for some variable $z_{m_j,p}^{(j)}$ to conclude that

$$Z = p^{1/2} \prod_{j=0}^{p-1} z_{m_j,p}^{(j)}.$$

It remains to find under which condition the identity (17) may be fulfilled. We write

$$\gamma_{(1+2j)/2p} \stackrel{(law)}{=} \gamma_{(1+2m_j)/2(m_j+1)}^{p/(m_j+1)} (z_{m_j,p}^{(j)})^{2p}. \tag{18}$$

Now, if $\frac{1+2j}{2p} < \frac{1+2m_j}{2(m_j+1)}$, we may apply the beta-gamma algebra to obtain

$$\gamma_{(1+2j)/2p} \stackrel{(law)}{=} \gamma_{(1+2m_j)/2(m_j+1)} \beta(\frac{1+2j}{2p}, \frac{1+2m_j}{2(m_j+1)} - \frac{1+2j}{2p})$$

but in (18), we need to have on the right-hand side $\gamma_{(1+2m_j)/2(m_j+1)}^{p/(m_j+1)}$ instead of $\gamma_{(1+2m_j)/2(m_j+1)}$.

However, it is known that

$$\gamma_a \stackrel{(law)}{=} \gamma_a^c \gamma_{a,c}$$

for some variable $\gamma_{a,c}$ independent of γ_a for any $c \in (0,1]$. This follows from the self-decomposable character of $\ln(\gamma_a)$. Thus, we seem to need $\frac{p}{m_j+1} \leq 1$. But, this condition is not compatible with (14) unless $m_j = m = p - 1$.

3.2 Asymptotic study

We study the behavior of the product $X_1^{(1)} ldots X_1^{(m+1)}$, resp. Z_m , appearing in the right-hand side of the equality in law (5), when $m \longrightarrow \infty$. Recall from (4) that

$$|X_1| \stackrel{(law)}{=} \left(2\gamma_{\frac{2m+1}{2(m+1)}}\right)^{\frac{1}{2(m+1)}}.$$

We are thus led to consider the product

$$\Theta_{a,b,c}^{(p)} = \left(\prod_{i=1}^p \gamma_{a-b/p}^{(i)}\right)^{c/p}$$

where in our set up of Theorem 1, p = m + 1, a = 1, b = c = 1/2.

$$\mathbb{E}[(\Theta_{a,b,c}^{(p)})^s] = \prod_{i=1}^p \mathbb{E}[\left(\gamma_{a-b/p}^{(i)}\right)^{cs/p}]$$

$$= \left(\frac{\Gamma(a - \frac{b}{p} + \frac{cs}{p})}{\Gamma(a - \frac{b}{p})}\right)^p$$

$$= \exp[p(\ln(\Gamma(a + \frac{cs - b}{p})) - \ln(\Gamma(a - \frac{b}{p})))]$$

$$\longrightarrow \exp(\frac{\Gamma'(a)}{\Gamma(a)}cs).$$

Thus, it follows that

$$\Theta_{a,b,c}^{(p)} \xrightarrow[p \to \infty]{\mathbb{P}} \exp(\frac{\Gamma'(a)}{\Gamma(a)}c),$$

implying that

$$|X_1^{(1)} \dots X_1^{(m+1)}| \underset{p \to \infty}{\overset{\mathbb{P}}{\longrightarrow}} \exp(-\gamma/2) \tag{19}$$

and

$$\exp(-\gamma/2)Z_m \underset{m \to \infty}{\overset{(law)}{\longrightarrow}} |N|. \tag{20}$$

where $\gamma = -\Gamma'(1)$ is the Euler constant.

We now look for a central limit theorem for $\Theta_{a,b,c}^{(p)}$. We consider the limiting distribution of

$$\sqrt{p} \left\{ \frac{c}{p} \sum_{i=1}^{p} \ln(\gamma_{a-b/p}^{(i)}) - c \frac{\Gamma'(a)}{\Gamma(a)} \right\}.$$

$$\mathbb{E}\left(\exp\left[cs\sqrt{p}\left\{\frac{1}{p}\sum_{i=1}^{p}\ln(\gamma_{a-b/p}^{(i)}) - \frac{\Gamma'(a)}{\Gamma(a)}\right\}\right]\right)$$

$$= \mathbb{E}\left[\prod_{i=1}^{p}\left(\gamma_{a-b/p}^{(i)}\right)^{cs/\sqrt{p}}\right] \exp(-cs\sqrt{p}\frac{\Gamma'(a)}{\Gamma(a)})$$

$$= \mathbb{E}\left[\left(\gamma_{a-b/p}^{(i)}\right)^{cs/\sqrt{p}}\right]^{p} \exp(-cs\sqrt{p}\frac{\Gamma'(a)}{\Gamma(a)})$$

$$= \left(\frac{\Gamma(a - \frac{b}{p} + \frac{cs}{\sqrt{p}})}{\Gamma(a - \frac{b}{p})}\right)^{p} \exp(-cs\sqrt{p}\frac{\Gamma'(a)}{\Gamma(a)})$$

$$= \exp[p(\ln(\Gamma(a - \frac{b}{p} + \frac{cs}{\sqrt{p}})) - \ln(\Gamma(a - \frac{b}{p}))) - cs\sqrt{p}\frac{\Gamma'(a)}{\Gamma(a)}]$$

$$= \exp(\frac{c^{2}s^{2}}{2}(\ln(\Gamma))''(a) + O(m^{-1/2}))$$

We thus obtain that

$$\sqrt{p} \left\{ \frac{c}{m} \sum_{i=1}^{m} \ln(\gamma_{a-b/m}^{(i)}) - c \frac{\Gamma'(a)}{\Gamma(a)} \right\} \stackrel{(law)}{\longrightarrow} N(0, \sigma^2)$$
 (21)

where $N(0, \sigma^2)$ denotes a centered Gaussian variable with variance:

$$\sigma^2 = c^2(\ln(\Gamma))''(a) = c^2 \left[\frac{\Gamma''(a)}{\Gamma(a)} - \left(\frac{\Gamma'(a)}{\Gamma(a)} \right)^2 \right].$$

or, equivalently

$$\left(\Theta_{a,b,c}^{(p)} \exp(\frac{\Gamma'(a)}{\Gamma(a)}c)\right)^{\sqrt{p}} \xrightarrow[p \to \infty]{(law)} \exp(N(0, c^2(\ln(\Gamma))''(a))). \tag{22}$$

References

 Albin, J. M. P. A continuous non-Brownian motion martingale with Brownian motion marginal distributions. Statist. Probab. Lett. 78 (2008), no. 6, 682–686.

- 2. Andrews, G.; Askey, R. and Roy, R. Special functions. Encyclopedia of Mathematics and its Applications, 71. Cambridge University Press, Cambridge, 1999.
- 3. Chaumont, L. and Yor, M. Exercises in probability. A guided tour from measure theory to random processes, via conditioning. Cambridge Series in Statistical and Probabilistic Mathematics, 13. Cambridge University Press, Cambridge, 2003.
- 4. Hamza, K. and Klebaner, F. A family of non-Gaussian martingales with Gaussian marginals. *J. Appl. Math. Stoch. Anal.* (2007), Art. ID 92723, 19 pp.
- Madan, D. and Yor, M. Making Markov martingales meet marginals: with explicit constructions. Bernoulli 8 (2002), no. 4, 509–536.