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Some properties of yarn

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## 1. Introduction

A yarn, of nonnegative cross-section, lies along a  $t$ -axis with  $-\infty < t < \infty$ . It is composed of fibres, which are cylinders with their axes parallel to the  $t$ -axis. The head of a fibre is its left endpoint on the  $t$ -axis, the tail its right endpoint. Fibre heads are distributed along the  $t$ -axis according to a Poisson-process with parameter  $\lambda(t)$ , where  $\lambda(t)$  is Lebesgue-integrable on any finite interval and  $0 < \lambda(t) \leq \lambda$  for all  $t$  and some finite  $\lambda$ . We define such a process in the following way: Regarding the  $t$ -axis as composed of intervals, of which the  $j^{\text{th}}$  interval is  $(j, j+1]$  (open on the left, closed on the right), where  $j$  runs through the integers, the distribution of the number  $\underline{r}_j$  of heads falling in the  $j^{\text{th}}$  interval is given by

$$(1.1) \quad P\{\underline{r}_j = r\} = e^{-\lambda_j} \frac{\lambda_j^r}{r!} \quad \text{for } r \geq 0$$

with

$$(1.2) \quad \lambda_j \stackrel{\text{def}}{=} \int_j^{j+1} \lambda(t) dt.$$

The head of the  $k^{\text{th}}$  fibre ( $k = 1, 2, \dots, r; \underline{r}_j = r \geq 1$ ) having its head in the  $j^{\text{th}}$  interval lies in  $\underline{t}_{j,k}$ , where under the condition  $\underline{r}_j = r$  the  $\underline{t}_{j,1}, \dots, \underline{t}_{j,r}$  are independently distributed random variables with the common distribution function  $K_j(t)$ , with

$$(1.3) \quad K_j(t) \stackrel{\text{def}}{=} F\{\underline{t}_{j,1} \leq t\} = \begin{cases} 0 & \text{for } t \leq j, \\ \lambda_j^{-1} \int_j^t \lambda(u) du & \text{for } j \leq t \leq j+1, \\ 1 & \text{for } j+1 \leq t. \end{cases}$$

The  $\underline{t}_{j,k}$  for different intervals on the  $t$ -axis are mutually independent.

The length  $\underline{x}$  ( $\geq 0$ ) and cross-section  $\underline{y}$  ( $\geq 0$ ) of a fibre have a simultaneous distribution function  $H(x,y) = P\{\underline{x} \leq x, \underline{y} \leq y\}$ . The vectors  $(\underline{x}, \underline{y})$  belonging to different fibres are mutually independent and independent of the location of the fibre on the  $t$ -axis. A realization of the yarn is a stepfunction, which can be taken continuous from the right. The length of the  $k^{\text{th}}$  fibre having its head in the  $j^{\text{th}}$  interval is  $\underline{x}_{j,k}$ , its cross-section  $\underline{y}_{j,k}$ .

Object of this paper is the study of random variables like  $\underline{c}(t_0)$  and  $\underline{v}(t_0, t_0+h)$ , which denote the cross-section of the yarn at  $t_0$  and the volume of yarn in the interval  $(t_0, t_0+h]$  respectively (with  $t_0$  arbitrary and  $h > 0$ ).

## 2. Terpstra's approach

In this section we assume that  $H(x_0, \infty) = 1$  for some  $x_0 < \infty$ , i.e. the fibres have length  $\leq x_0$  with probability 1. Furthermore we take  $\lambda(t) = \lambda > 0$ , i.e. we consider a stationary Poisson-process (this restriction need not be made).

In order to derive  $\mathcal{E} e^{i\tau \underline{c}(t_0)}$ , we first compute

$$(2.1) \quad \varphi_{\underline{c}}(\tau, t_0; s, 1) \stackrel{\text{def}}{=} \mathcal{E} \{ e^{i\tau \underline{c}(t_0)} \mid \underline{n}(s, s+1) = 1 \},$$

where  $\underline{n}(s, s+1)$  denotes the number of heads of fibres in the interval  $(s, s+1]$  and  $s < t_0 - x_0 < t_0 < s+1$ .

If  $\underline{t}$  is the coordinate of the head of the one fibre falling in  $(s, s+1]$ , then it is known that  $\underline{t}$  is uniformly distributed in this interval. If the fibre has length  $\underline{x} = x$ , with probability  $\frac{x}{1}$  the head lies in  $(t_0 - x, t_0]$ . Hence

$$(2.2) \quad \begin{aligned} \varphi_{\underline{c}}(\tau, t_0; s, 1) &= \int_0^\infty \int_0^\infty \left[ e^{i\tau y} \frac{x}{1} + e^0 \left(1 - \frac{x}{1}\right) \right] dH(x, y) = \\ &= 1 - \frac{1}{1} \int_0^\infty \int_0^\infty x(1 - e^{i\tau y}) dH(x, y). \end{aligned}$$

As

$$(2.3) \quad P\{\underline{n}(s, s+1) = n\} = e^{-\lambda 1} \frac{(\lambda 1)^n}{n!} \quad \text{for } n = 0, 1, 2, \dots$$

and  $\underline{c}(t_0)$  is the sum of the independent contributions to the cross-section of all fibres with heads in  $(s, s+1]$ , we have (the characteristic function of a sum of independent random variables being equal to the product of the characteristic functions of the individual terms)

$$(2.4) \quad \begin{aligned} \varphi_{\underline{c}}(\tau, t_0) &\stackrel{\text{def}}{=} \mathcal{E} e^{i\tau \underline{c}(t_0)} = \\ &= \sum_{n=0}^{\infty} \{ \varphi_{\underline{c}}(\tau, t_0; s, 1) \}^n e^{-\lambda 1} \frac{(\lambda 1)^n}{n!} = \\ &= \exp -\lambda 1 \left\{ 1 - \left(1 - \frac{1}{1} \int_0^\infty \int_0^\infty x(1 - e^{i\tau y}) dH(x, y) \right) \right\} = \\ &= \exp -\lambda \int_0^\infty \int_0^\infty x(1 - e^{i\tau y}) dH(x, y). \end{aligned}$$

As was to be expected,  $\varphi_{\underline{c}}(\tau, t_0)$  does not depend on either  $s$  or  $1$ . We may conjecture, that (2.4) will also hold if  $H(x, \infty) < 1$  for all  $x < \infty$ . (Some restriction is needed: we must have  $\mathcal{E} \underline{x} < \infty$ , cf. Breny (1957) and section 4.)

We remark that from (2.4) it is evident that  $\underline{c}(t_0)$  has a compound Poisson - distribution, i.e.

$$(2.5) \quad \underline{c}(t_0) = \sum_{j=1}^n \underline{b}_j,$$

where

$$(2.6) \quad P\{\underline{n} = n\} = e^{-\lambda \mathcal{E} \underline{x}} \frac{(\lambda \mathcal{E} \underline{x})^n}{n!} \quad \text{for } n = 0, 1, 2, \dots$$

and, under the condition  $\underline{n} = n$ , the  $\underline{b}_j$  are independently distributed, each with distribution function

$$(2.7) \quad G(y) \stackrel{\text{def}}{=} \frac{\int_0^\infty x d_x H(x, y)}{\mathcal{E} \underline{x}}.$$

Analogously to the derivation of (2.4) one can show that

$$(2.8) \quad \varphi_{\underline{y}}(\tau, t_0, h) \stackrel{\text{def}}{=} \mathcal{E} e^{i\tau \underline{y}(t_0, t_0+h)} =$$

$$= \exp - \lambda [h + \mathcal{E} \underline{x} + \int_{y=0}^\infty \int_{x=0}^h \{ \frac{2}{iy\tau} (1 - e^{i\tau xy}) - (h-x) e^{i\tau xy} \} dH(x, y) +$$

$$+ \int_{y=0}^\infty \int_{x=h}^\infty \frac{2}{iy\tau} (1 - e^{i\tau hy}) - (x-h) e^{i\tau hy} dH(x, y)],$$

always assuming a finite maximum length  $x_0$  for the fibres.

Both (2.4) and (2.8) are due to Terpstra. These results are known for  $\underline{y} = 1$  with probability 1 (cf. Spencer-Smith and Todd (1941), Martindale (1945), Breny (1952) and Breny (1953), Olerup (1952)). Related considerations as to method of derivation are to be found in Fortet (1951).

### 3. Van Dantzig's method

In this section we assume that  $\underline{y} \equiv 1$  (or  $\underline{y} = 1$  with probability 1).

In Breny (1957) relation (2.8) (with  $\underline{y} \equiv 1$ ) is obtained by applying a limiting procedure to (in our notation)  $\mathcal{E} \exp i \sum_{j=0}^n \tau_j \underline{n}(t_j)$ , expressed as a complicated sum of double integrals. As Van Dantzig pointed out to Breny (cf. Breny (1957), page 33) one might proceed in the following manner. Let  $A$  denote a Lebesgue-measurable set in the space

$\Omega = \{(t, \underline{x}) \mid -\infty < t < \infty, 0 \leq x < \infty\}$  and  $\underline{m}(A)$  the number of fibres for which  $(\underline{t}_{j,k}, \underline{x}_{j,k}) \in A$  is satisfied. Then, if  $A \cap B = \emptyset$ ,  $\underline{m}(A)$  and  $\underline{m}(B)$

are independent stochastic variables having a Poisson - distribution, the parameter of  $\underline{m}(A)$  being

$$(3.1) \quad \xi_{\underline{m}(A)} = \iint_A \lambda(t) dt dF(x),$$

where  $F(x) \stackrel{\text{def}}{=} P\{x \leq \lambda\}$  is the distribution function of the length of a fibre. For a general class of real functions  $\xi(t, x)$  one can define

$$(3.2) \quad \int_0^\infty \int_{-\infty}^\infty \xi(t, x) d\underline{m},$$

where  $\underline{m}$  is a stochastic measure on  $\Omega$ , with  $\underline{m}(A)$  as described, for every L-measurable set A. Because of the independence of the  $\underline{m}(A)$  for disjoint sets, we shall have something like (use Riemann - sums)

$$(3.3) \quad \begin{aligned} \mathcal{E} \exp i\tau \int_0^\infty \int_{-\infty}^\infty \xi(t, x) d\underline{m} &\approx \prod_{\nu=0}^\infty \prod_{\mu=-\infty}^\infty \mathcal{E} e^{i\tau \xi(t_\mu^*, x_\nu^*) \underline{m}(\Delta_{\mu\nu})} \approx \\ &\approx \prod_{\nu=0}^\infty \prod_{\mu=-\infty}^\infty \left\{ (1 - \lambda(t) \Delta t \Delta F(x)) e^0 + \lambda(t) \Delta t \Delta F(x) e^{i\tau \xi(t_\mu^*, x_\nu^*)} \right\} \approx \\ &\approx \prod_{\nu=0}^\infty \prod_{\mu=-\infty}^\infty \exp - \lambda(t) \{1 - i\tau \xi(t_\mu^*, x_\nu^*)\} \Delta t \Delta F(x) \approx \\ &\approx \exp - \int_0^\infty \int_{-\infty}^\infty \lambda(t) \{1 - i\tau \xi(t, x)\} dt dF(x) \end{aligned}$$

or

$$(3.4) \quad \mathcal{E} \exp i\tau \int_0^\infty \int_{-\infty}^\infty \xi(t, x) d\underline{m} = \exp - \int_0^\infty \int_{-\infty}^\infty \lambda(t) \{1 - i\tau \xi(t, x)\} dt dF(x).$$

Substitution of

$$(3.5) \quad \xi(t, x) = \begin{cases} 1 & \text{for } t \leq t_0, t+x > t_0 \\ 0 & \text{otherwise} \end{cases},$$

now leads to (2.4), because

$$(3.6) \quad \underline{c}(t_0) = \int_0^\infty \int_{t_0-x < t \leq t_0} d\underline{m}.$$

In the same way

$$(3.7) \quad \begin{aligned} \xi(t, x) &= \{(t-t_0) \wedge (t_0-t) - (t-t_0-h) \wedge (t_0+h-t)\} + \\ &- \{(x+t-t_0) \wedge (t_0-x-t) - (x+t-t_0-h) \wedge (t_0+h-x-t)\} \end{aligned}$$

leads to (2.8).

Stochastic set functions have been discussed by Prékopa (cf. Prékopa (1956, 1957)).

#### 4. Main formula

The  $k^{\text{th}}$  fibre from the  $j^{\text{th}}$  interval contributes to the cross-section at point  $t$  of the axis an amount

$$(4.1) \quad \underline{c}_{j,k}(t) \stackrel{\text{def}}{=} y_{j,k} \{ \iota(\underline{t}_{j,k} + \underline{x}_{j,k} - t) - \iota(\underline{t}_{j,k} - t) \},$$

where  $\iota(x) = 1$  for  $x \geq 0$  and  $\iota(x) = 0$  for  $x < 0$ . Hence we find for the total cross-section at point  $t$  <sup>1)</sup>

$$(4.2) \quad \underline{c}(t) \stackrel{\text{def}}{=} \sum_{j=-\infty}^{\infty} \sum_{k=1}^{r_j} y_{j,k} \{ \iota(\underline{t}_{j,k} + \underline{x}_{j,k} - t) - \iota(\underline{t}_{j,k} - t) \}.$$

We shall study the random variable

$$(4.3) \quad \underline{u}_T \stackrel{\text{def}}{=} \int_{-\infty}^{\infty} \underline{c}(t) dT(t)$$

under the condition  $\sum \underline{x} < \infty$ . This variable is (as will be shown) well-defined for any real-valued function  $T(t)$ , which is of bounded variation in a closed interval  $[t_1, t_2]$  (with  $-\infty < t_1 < t_2 < \infty$ ), constant for  $t \leq t_1$ , as well as constant for  $t \geq t_2$  and finally continuous from the right for all  $t$ . We assume (this is no restriction) that  $t_1$  is a negative integer and  $t_2$  a positive integer.

With probability 1  $\underline{c}(t)$  is a stepfunction with a finite number of finite steps in the interval  $[t_1, t_2]$ , given  $\sum \underline{x} < \infty$  (Breny (1957)). As inside  $[t_1, t_2]$  only a finite number of finite steps originate, it is sufficient to prove that with probability 1 only a finite number of fibres cover the point  $t_1$ . Thus we have to prove

$$(4.4) \quad P \left\{ \sum_{j=-\infty}^{t_1-1} \sum_{k=1}^{r_j} \iota(\underline{t}_{j,k} + \underline{x}_{j,k} - t_1) < \infty \right\} = 1$$

or

1) We take  $\sum_{k=1}^{r_j} \dots = 0$  for  $r_j = 0$ , here and later.

$$(4.5) \quad P \left\{ \max_{1 \leq k \leq r_j} (\underline{t}_{j,k} + \underline{x}_{j,k} - t_1) \geq 0 \text{ for infinitely many } j \right\} = 0.$$

To prove the last relation, it is sufficient to show (by one of the Borel - Cantelli lemmas) that for the (independent) events

$$(4.6) \quad A_j \stackrel{\text{def}}{=} \left\{ \max_{1 \leq k \leq r_j} (\underline{t}_{j,k} + \underline{x}_{j,k} - t_1) \geq 0 \right\}$$

we have

$$(4.7) \quad \sum_{j=-\infty}^{t_1-1} P\{A_j\} < \infty.$$

Now, if  $F(x) \stackrel{\text{def}}{=}} P\{\underline{x} \leq x\}$ ,

$$(4.8) \quad \begin{aligned} P\{A_j\} &= P \left\{ \max_{1 \leq k \leq r_j} (\underline{t}_{j,k} + \underline{x}_{j,k} - t_1) \geq 0 \right\} = \\ &= \sum_{r=1}^{\infty} e^{-\lambda_j} \frac{\lambda_j^r}{r!} \left\{ 1 - \left[ \int_j^{j+1} F(t_1 - t) dK_j(t) \right]^r \right\} = \\ &= 1 - \exp - \lambda_j \left\{ 1 - \int_j^{j+1} F(t_1 - t) dK_j(t) \right\} \leq \\ &\leq \lambda_j \left\{ 1 - \int_j^{j+1} F(t_1 - t) dK_j(t) \right\} \leq \lambda \{1 - F(t_1 - j - 1)\}, \end{aligned}$$

as  $1 - e^{-x} \leq x$  for  $x \geq 0$  and  $\lambda_j \leq \lambda$  for each  $j$ . Hence, because  $\sum_{k=1}^{\infty} \{1 - F(k)\} \leq \int_0^{\infty} x dF(x) = \mathcal{E}\underline{x} < \infty$ , condition (4.7) is satisfied.

For a stochastic stepfunction  $\underline{c}(t)$  which has, with probability 1, a finite number of finite steps in the interval  $[t_1, t_2]$ , the integral in (4.3) is defined with probability 1 as the Lebesgue - Stieltjes integral of the realization  $c(t)$  and satisfies

$$(4.9) \quad \begin{aligned} \underline{u}_T &= \int_{-\infty}^{\infty} \underline{c}(t) dT(t) = \\ &= \sum_{j=-\infty}^{\infty} \sum_{k=1}^{r_j} \underline{y}_{j,k} \int_{-\infty}^{\infty} \{L(\underline{t}_{j,k} + \underline{x}_{j,k} - t) - L(\underline{t}_{j,k} - t)\} dT(t) = \\ &= \sum_{j=-\infty}^{\infty} \sum_{k=1}^{r_j} \underline{y}_{j,k} \{T(\underline{t}_{j,k} + \underline{x}_{j,k}) - T(\underline{t}_{j,k})\}, \end{aligned}$$

where the double series has only a finite number of nontrivial terms (i.e. terms unequal 0) with probability 1.

Lemma. If  $\underline{a}_n$ ,  $n = 1, 2, \dots$  are mutually independent real-valued stochastic variables, such that the events  $A_n = \{\underline{a}_n \neq 0\}$  satisfy  $P\{A_n \text{ occurs for infinitely many } n\} = 0$ , we have:  $\underline{s} \stackrel{\text{def}}{=} \sum_{n=1}^{\infty} \underline{a}_n$  is a stochastic variable for which (for all real  $\tau$ )

$$(4.10) \quad \mathcal{E} e^{i\tau \underline{s}} = \prod_{n=1}^{\infty} \mathcal{E} e^{i\tau \underline{a}_n}.$$

Proof:  $\underline{s}$  is a well-defined stochastic variable, for it is with probability 1 the sum of a finite number of independent random variables unequal 0. As we have

$$(4.11) \quad \{A_n \text{ occurs for infinitely many } n\} = \bigcap_{n=1}^{\infty} \bigcup_{m=n}^{\infty} A_m$$

and therefore

$$(4.12) \quad \lim_{n \rightarrow \infty} P\{\bigcup_{m=n}^{\infty} A_m\} = 0,$$

to each  $\varepsilon > 0$  there exists an  $N = N(\varepsilon)$ , for which  $P\{\bigcup_{m=N+1}^{\infty} A_m\} < \varepsilon$  or  $P\{\bigcap_{m=N+1}^{\infty} \bar{A}_m\} > 1 - \varepsilon$  (if  $\bar{A}_m$  denotes the complement of the set  $A_m$ ). But then

$$(4.13) \quad \left| \mathcal{E} \prod_{n=1}^{\infty} e^{i\tau \underline{a}_n} - \mathcal{E} \prod_{n=1}^N e^{i\tau \underline{a}_n} \right| \leq \left| \mathcal{E} \prod_{n=1}^N e^{i\tau \underline{a}_n} \right| \cdot \left| \mathcal{E} \prod_{n=N+1}^{\infty} e^{i\tau \underline{a}_n} - 1 \right| \leq \\ \leq \mathcal{E} \left\{ \left| \prod_{n=N+1}^{\infty} e^{i\tau \underline{a}_n} - 1 \right| \middle| a_n \neq 0 \text{ for at least one } n \geq N+1 \right\} \cdot \varepsilon \leq 2\varepsilon,$$

and hence

$$(4.14) \quad \lim_{N \rightarrow \infty} \prod_{n=1}^N \mathcal{E} e^{i\tau \underline{a}_n} = \mathcal{E} \prod_{n=1}^{\infty} e^{i\tau \underline{a}_n},$$

which proves the lemma.

The random variables

$$(4.15) \quad \underline{z}_j \stackrel{\text{def}}{=} \prod_{k=1}^{r_j} \exp i\tau Y_{j,k} \{T(\underline{t}_{j,k} + \underline{x}_{j,k}) - T(\underline{t}_{j,k})\}$$

are nontrivial, i.e. not equal to 1 with probability 1, only for  $j \leq t_2 - 1$ . Hence we find from (4.9) by applying the lemma to

$$(4.16) \quad \underline{a}_n \stackrel{\text{def}}{=} \underline{z}_{t_2-n},$$



that for all real  $\tau$

$$\begin{aligned}
 (4.17) \quad \mathcal{E} \exp i\tau \underline{u}_T &= \\
 &= \mathcal{E} \prod_{j=-\infty}^{\infty} \prod_{k=1}^{r_j} \exp i\tau \underline{y}_{j,k} \{T(\underline{t}_{j,k} + \underline{x}_{j,k}) - T(\underline{t}_{j,k})\} = \\
 &= \prod_{j=-\infty}^{\infty} \mathcal{E} \prod_{k=1}^{r_j} \exp i\tau \underline{y}_{j,k} \{T(\underline{t}_{j,k} + \underline{x}_{j,k}) - T(\underline{t}_{j,k})\}.
 \end{aligned}$$

Further we have by (1.1), (1.2) and (1.3), using Fubini's theorem to prove the second equality,

$$\begin{aligned}
 (4.18) \quad \mathcal{E} \prod_{k=1}^{r_j} \exp i\tau \underline{y}_{j,k} \{T(\underline{t}_{j,k} + \underline{x}_{j,k}) - T(\underline{t}_{j,k})\} &= \\
 &= \mathcal{E} \sum_{r=0}^{\infty} \left\{ \prod_{k=1}^r \exp i\tau \underline{y}_{j,k} [T(\underline{t}_{j,k} + \underline{x}_{j,k}) - T(\underline{t}_{j,k})] \Big|_{\underline{r}_j = r} \right\} P\{\underline{r}_j = r\} = \\
 &= \sum_{r=0}^{\infty} \left\{ \int_0^{\infty} \int_0^{\infty} \int_j^{j+1} \exp i\tau \underline{y} [T(t+x) - T(t)] dK_j(t) dH(x,y) \right\}^r P\{\underline{r}_j = r\} = \\
 &= \exp \left\{ -\lambda_j \left( 1 - \int_0^{\infty} \int_0^{\infty} \int_j^{j+1} \exp i\tau \underline{y} [T(t+x) - T(t)] dK_j(t) dH(x,y) \right) \right\} = \\
 &= \exp - \int_0^{\infty} \int_0^{\infty} \int_j^{j+1} \lambda(t) [1 - \exp i\tau \underline{y} \{T(t+x) - T(t)\}] dt dH(x,y).
 \end{aligned}$$

Because  $|1 - \exp i\tau \underline{y} \{T(t+x) - T(t)\}| \leq 2$  and  $T(t+x) - T(t) = 0$  for  $t \leq t_1 - x$  and  $t \geq t_2$ , we have

$$(4.19) \quad \int_{-\infty}^{\infty} |1 - \exp i\tau \underline{y} \{T(t+x) - T(t)\}| dt \leq 2\{x + (t_2 - t_1)\}$$

and so, because

$$(4.20) \quad \int_0^{\infty} \int_0^{\infty} \int_{-\infty}^{\infty} |1 - \exp i\tau \underline{y} \{T(t+x) - T(t)\}| dt dH(x,y) \leq 2\{\mathcal{E} \underline{x} + (t_2 - t_1)\} < \infty,$$

we are allowed to apply Fubini's theorem once again. Combining (4.17) and (4.18) we find our main formula

$$\begin{aligned}
 (4.21) \quad \mathcal{E} \exp i\tau \underline{u}_T &= \\
 &= \exp - \sum_{j=-\infty}^{\infty} \int_0^{\infty} \int_0^{\infty} \int_j^{j+1} \lambda(t) [1 - \exp i\tau \underline{y} \{T(t+x) - T(t)\}] dt dH(x,y) = \\
 &= \exp - \int_0^{\infty} \int_0^{\infty} \int_{-\infty}^{\infty} \lambda(t) [1 - \exp i\tau \underline{y} \{T(t+x) - T(t)\}] dt dH(x,y).
 \end{aligned}$$

The particular cases

$$(4.22) \quad T(t) = L(t - t_0),$$

$$(4.23) \quad T(t) = (t - t_0)L(t - t_0) - (t - t_0 - h)L(t - t_0 - h),$$

show, that both (2.4) and (2.8) are satisfied for  $E_x < \infty$ . In fact a generalization of these results (with  $\lambda(t)$  instead of  $\lambda$ ) has been obtained.

Relation (4.21) may be used to obtain interesting formulae for other stochastic variables besides  $c(t_0)$  and  $v(t_0, t_0 + h)$ .

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