

## MOMENT MEASURES OF MIXED EMPIRICAL RANDOM POINT PROCESSES AND MARKED POINT PROCESSES IN COMPACT METRIC SPACES. 2

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ABSTRACT. This is a continuation of the paper by M. G. Semeïko, *Moment measures of mixed empirical random point processes and marked point processes in compact metric spaces. I*, Theor. Probability and Math. Statist. **88** (2014), 161–174. Moment measures of mixed empirical marked random point processes are investigated by using the probability generating functions of random counting measures.

This is a continuation of the paper [1]. The numbering of equations in the current paper continues the numbering in [1].

### 7. MAIN DEFINITIONS OF THE THEORY OF RANDOM MIXED EMPIRICAL ORDERED MARKED POINT PROCESSES

Every trajectory  $E^*$  of a finite strictly simple ordered marked point process

$$\mathcal{D} = (\mathcal{E}^*, \mathfrak{X}^*, P^*)$$

in a bounded space ( $Y = X \times K, \mathfrak{A}_Y = \mathfrak{A}_X \otimes \mathfrak{A}_K, \mathcal{B}_Y = \mathcal{B}_X \odot \mathcal{B}_K$ ) is a thinned set in the Cartesian product  $Y = X \times K$  [2]. If  $X$  is a compact metric space of states endowed with a measure  $\vartheta$ , metric  $\rho_X(x_i, x_j)$ , and natural structures of measurable sets  $\mathfrak{A}_X$  and bounded sets  $\mathfrak{B}_X$  and if the space of marks  $K$  is an interval  $[a, b] \subset R^1$ , then every trajectory  $E^*$  of an ordered marked point process consists of a finite sequence of points:  $E^* = (y_1, \dots, y_i, \dots, y_n) = ([x_1; k_1], \dots, [x_i; k_i], \dots, [x_n; k_n])$ , where  $y_i = [x_i; k_i]$ ,  $x_i$  is a state and  $k_i$  is its mark. The phase space  $Y = X \times K$  can be endowed with the structure of a metric space by defining the distance between the points  $y_i = [x_i; k_i]$  and  $y_j = [x_j; k_j]$ ,  $i \neq j$ , by

$$\rho_Y([x_i; k_i], [x_j; k_j]) = \rho_X(x_i, x_j) + |k_i - k_j|.$$

Consider the following random procedure of constructing an ordered marked point process. Introduce the following three random variables:  $x = x(\omega)$ ,  $k = k(\omega)$ , and  $\nu = \nu(\omega)$  and assume that these random variables satisfy the following conditions.

7.1. The random variables  $x(\omega)$ ,  $k(\omega)$ , and integer valued nonnegative random variable  $\nu(\omega)$  are defined on the main probability space  $(\Omega, \mathfrak{F}, P)$ .

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7.2. The random variables  $x(\omega)$ ,  $k(\omega)$ , and  $\nu(\omega)$  assume values in the sample spaces  $(X, \mathfrak{A}_X, P_x)$ ,  $(K, \mathfrak{A}_K, P_k)$ , and  $(Z_+, \mathfrak{A}_{Z_+}, P_\nu)$ , respectively, where

$$\begin{aligned} P_x(B_X) &= \mathbb{P}\{\omega: x(\omega) \in B_X\} = \mu_1(B_X), & B_X \in \mathfrak{A}_X, \\ P_k(B_K) &= \mathbb{P}\{\omega: k(\omega) \in B_K\} = \mu_2(B_K), & B_K \in \mathfrak{A}_K, \end{aligned}$$

and

$$P_\nu(B_{Z_+}) = \mathbb{P}\{\omega: \nu(\omega) \in B_{Z_+}\}, \quad B_{Z_+} \in \mathfrak{A}_{Z_+}.$$

7.3. The distribution  $P_x(B_X)$  of the random variable  $x = x(\omega)$  is absolutely continuous with respect to the measure  $\vartheta$  in the measurable space  $(X, \mathfrak{A}_X)$ .

7.4. The distribution  $P_k(B_K)$  of the random variable  $k = k(\omega)$  is absolutely continuous with respect to the Lebesgue measure in the measurable space  $(\mathbf{R}^1, \mathfrak{A}_{\mathbf{R}^1})$ .

7.5.  $x(\omega)$ ,  $k(\omega)$ , and  $\nu(\omega)$  are jointly independent random variables.

Consider the product of probability measures  $P_y = P_x \otimes P_k$  on the  $\sigma$ -algebra of Borel sets  $\mathfrak{A}_Y = \mathfrak{A}_X \otimes \mathfrak{A}_K$  of the phase space  $Y = X \times K$ . Then  $(Y, \mathfrak{A}_Y, P_y)$  can be viewed as a sample probability space for the two dimensional random variable  $y(\omega) = [x(\omega); k(\omega)]$  that generates the probability measure

$$\begin{aligned} P_y(B_Y) &= P_y(B_X \times B_K) = \mathbb{P}\{\omega: [x(\omega); k(\omega)] \in B_X \times B_K\} \\ &= \mathbb{P}\{\omega: x(\omega) \in B_X, k(\omega) \in B_K\} = \mathbb{P}\{\omega: x(\omega) \in B_X\} \mathbb{P}\{\omega: k(\omega) \in B_K\} \\ &= P_x(B_X)P_k(B_K) = \mu_1(B_X)\mu_2(B_K) = \mu(B_X \times B_K) = \mu(B_Y), \end{aligned}$$

where  $B_Y = B_X \times B_K \in \mathfrak{E}_Y = \mathfrak{A}_Y \cap \mathcal{B}_Y$ .

Let  $G_1$  and  $G_2$  be two independent random experiments corresponding to the sample probability spaces  $(X, \mathfrak{A}_X, P_x)$  and  $(K, \mathfrak{A}_K, P_k)$ . Then  $G = (G_1, G_2)$  is a ‘‘compound’’ random experiment with the sample probability space  $(Y, \mathfrak{A}_Y, P_y)$ . A number  $n \in Z_+$  is chosen at random and then every trajectory of the ordered marked point process

$$E^* = ([x_1; k_1], \dots, [x_i; k_i], \dots, [x_n; k_n])$$

of size  $n$  is formed as a result of  $n$  independent repetitions of the same ‘‘compound’’ random experiment  $G = (G_1, G_2)$  that can be described as a simple random choice without repetition of a pair  $y_i = [x_i; k_i]$ ,  $i = 1, \dots, n$ , from the phase space  $Y = X \times K$ : the state point  $x_i$  is chosen from the space  $X$  (experiment  $G_1$ ), while its mark  $k_i$  is chosen from the space  $K$  (experiment  $G_2$ ).

Thus a trajectory  $E^*$  can be viewed as a realization in the sample measurable space  $(Y, \mathfrak{A}_Y)$  of the following sequence:

$$\begin{aligned} E^* &= E^*(\omega) = (y_1(\omega), \dots, y_i(\omega), \dots, y_{\nu(\omega)}(\omega)) \\ &= ([x_1(\omega); k_1(\omega)], \dots, [x_i(\omega); k_i(\omega)], \dots, [x_{\nu(\omega)}(\omega); k_{\nu(\omega)}(\omega)]) \end{aligned}$$

of a random size  $\nu(\omega)$  of independent and identically distributed random elements (two dimensional random variables) defined in the main probability space  $(\Omega, \mathfrak{F}, \mathbb{P})$  and distributed according to the probability measure  $P_y$ , where  $y_i(\omega) = [x_i(\omega); k_i(\omega)]$ .

It is clear that the random variable  $\nu(\omega)$  admits the following representation:

$$\nu(\omega) = N^* = N^*(E^*, Y) = \text{card}[E^* \cap Y] = \sum_{y \in E^*} I_Y(y),$$

where  $N^*(E^*, Y)$  is the random variable determining the number of points in the set  $E^*$  of the space  $Y$  and where  $I_Y(y)$  is the characteristic function of the space  $Y$ .

**Definition 7.1.** A random process  $\mathcal{D} = (\mathcal{E}^*, \mathfrak{X}^*, P^*)$  is called a strictly simple mixed empirical ordered marked point process with independent marks in a bounded space  $(Y, \mathfrak{A}_Y, \mathcal{B}_Y)$  [2, 3].

Given  $n$ ,  $N^* = n$ , and an arbitrary bounded measurable set  $B_Y = B_X \times B_K$ ,  $B_X \in \mathfrak{E}_X$ ,  $B_K \in \mathfrak{A}_K$ , consider a random empirical counting measure of the ordered marked point process

$$N^*(B_Y) = N^*(E^*, B_Y) = \text{card}[E^* \cap B_Y] = \sum_{i=1}^n I_{B_Y}([x_i; k_i]).$$

The counting measure  $N^*(B_Y)$  has the binomial distribution  $B(n, \mu(B_Y))$  with parametric measure  $\mu(B_Y)$  [3]:

$$(34) \quad \begin{aligned} P^*\{N^*(B_Y) = k \mid N^* = n\} &= C_n^k \mu^k(B_Y) [1 - \mu(B_Y)]^{n-k} \\ &= C_n^k \mu_1^k(B_X) \mu_2^k(B_K) [1 - \mu_1(B_X) \mu_2(B_K)]^{n-k}, \\ &k = 0, 1, \dots, n. \end{aligned}$$

If  $\{B_Y^j = B_X^j \times B_K^j : j = 1, \dots, s, s \geq 2, \bigcup_{j=1}^s B_Y^j = Y\}$  is an arbitrary finite sequence of disjoint bounded measurable sets of the phase space  $Y$ ,  $k_j \in \mathbb{Z}_+$ ,  $j = 1, \dots, s$ ,  $k_1 + k_2 + \dots + k_s = n$ , then, given  $N^* = n$ , the joint conditional distribution of counting measures  $\{N^*(B_Y^j), j = 1, \dots, s\}$  has the polynomial distribution [4]:

$$(35) \quad P^*\left\{N^*(B_Y^j) = k_j, j = 1, \dots, s \mid N^* = n\right\} = \frac{n!}{k_1! \dots k_s!} \mu^{k_1}(B_Y^1) \dots \mu^{k_s}(B_Y^s),$$

where  $\sum_{j=1}^s \mu(B_Y^j) = 1$ .

## 8. MOMENT MEASURES OF A MIXED EMPIRICAL RANDOM ORDERED MARKED POINT PROCESS WITH INDEPENDENT MARKS

We construct the joint probability generating function of the counting measures

$$\left\{N^*(B_Y^j), j = 1, \dots, s\right\},$$

where  $\bigcup_{j=1}^s B_Y^j = Y(B_Y^i \cap B_Y^r = \emptyset, 1 \leq i, r \leq s, i \neq r)$ , by using the joint conditional distribution (35) and a similar evaluation presented in Section 4 of [1]:

$$(36) \quad \begin{aligned} \Pi_{N^*(B_Y^1), \dots, N^*(B_Y^s)}(z_1, \dots, z_s) &= \Pi_{N^*}(z_1 \mu(B_Y^1) + \dots + z_s \mu(B_Y^s)) \\ &= \Pi_{N^*}(z_1 \mu_1(B_X^1) \mu_2(B_K^1) + \dots + z_s \mu_1(B_X^s) \mu_2(B_K^s)). \end{aligned}$$

Introduce the following notation for the measures of an ordered marked point process  $\mathcal{D}$ :

1.  $\nu_{\mathcal{D}}^{(h)}(B_Y^j) = M[\{N^*(B_Y^j)\}^h]$  is the moment measure of order  $h$ ,  $h = 1, 2, \dots$ ;
2.  $\nu_{\mathcal{D}}^{(h)}(B_Y^{j_1} \times \dots \times B_Y^{j_h}) = M[N^*(B_Y^{j_1}) \dots N^*(B_Y^{j_h})]$  is the mixed moment measure of order  $h$ ,  $1 \leq j_1 < \dots < j_h \leq s$ ,  $h = 1, \dots, s$ ;
3.  $\alpha_{\mathcal{D}}^{(h)}(B_Y^j) = M[N^*(B_Y^j)(N^*(B_Y^j) - 1) \dots (N^*(B_Y^j) - h + 1)]$  is the factorial moment measure of order  $h$ ;
4.  $\nu_{\mathcal{D}}^{(2)}(B_Y^{j_1} \times B_Y^{j_2}) = M[N^*(B_Y^{j_1}) N^*(B_Y^{j_2})]$  is the mixed moment measure of the second order,  $1 \leq j_1, j_2 \leq s$ ,  $j_1 \neq j_2$ ;
5.  $D(N^*(B_Y^j)) = \nu_{\mathcal{D}}^{(2)}(B_Y^j) - \{\nu_{\mathcal{D}}^{(1)}(B_Y^j)\}^2$  is the variance of the counting measure  $N^*(B_Y^j)$ ;
6.  $\text{cov}[N^*(B_Y^{j_1}), N^*(B_Y^{j_2})] = \nu_{\mathcal{D}}^{(2)}(B_Y^{j_1} \times B_Y^{j_2}) - \nu_{\mathcal{D}}^{(1)}(B_Y^{j_1}) \nu_{\mathcal{D}}^{(1)}(B_Y^{j_2})$  is the covariance measure of dependence between the measures  $N^*(B_Y^{j_1})$  and  $N^*(B_Y^{j_2})$ .

Reasoning as in Section 4 of [1], we obtain

$$(37) \quad \nu_{\mathcal{D}}^{(h)}(B_Y^{j_1} \times \cdots \times B_Y^{j_h}) = \mu(B_Y^{j_1}) \cdots \mu(B_Y^{j_h}) \Pi_{N^*}^{(h)}(1),$$

$$(38) \quad \alpha_{\mathcal{D}}^{(h)}(B_Y^j) = \mu^h(B_Y^j) \Pi_{N^*}^{(h)}(1), \quad \nu_{\mathcal{D}}^{(1)}(B_Y^j) = \mu(B_Y^j) \Pi_{N^*}'(1),$$

$$(39) \quad \nu_{\mathcal{D}}^{(2)}(B_Y^j) = \mu^2(B_Y^j) \Pi_{N^*}''(1) + \mu(B_Y^j) \Pi_{N^*}'(1),$$

$$(40) \quad \nu_{\mathcal{D}}^{(2)}(B_Y^{j_1} \times B_Y^{j_2}) = \mu(B_Y^{j_1}) \mu(B_Y^{j_2}) \Pi_{N^*}''(1),$$

$$(41) \quad D(N^*(B_Y^j)) = \mu^2(B_Y^j) [\Pi_{N^*}''(1) - \{\Pi_{N^*}'(1)\}^2] + \mu(B_Y^j) \Pi_{N^*}'(1),$$

$$(42) \quad \text{cov}[N^*(B_Y^{j_1}), N^*(B_Y^{j_2})] = \mu(B_Y^{j_1}) \mu(B_Y^{j_2}) [\Pi_{N^*}''(1) - \{\Pi_{N^*}'(1)\}^2].$$

## 9. MIXED EMPIRICAL POISSON RANDOM ORDERED MARKED POINT PROCESS WITH INDEPENDENT MARKS

### Theorem 9.1. *If*

1. *the size of a sample  $\nu = \nu(\omega) = N^*$  is a random variable with the Poisson distribution with parameter  $\lambda$ ;*
2. *the random variable  $\nu$  is independent of the random variables*

$$\{[x_i(\omega); k_i(\omega)]: i = 1, \dots, \nu(\omega)\};$$

3.  *$N^*(B_Y) = \sum_{i=1}^{\nu} I_{B_Y}([x_i; k_i])$  is the random empirical counting measure of the ordered marked point process;*
4.  *$\{B_Y^j = B_X^j \times B_K^j: j = 1, \dots, s, s \geq 2\}$  is an arbitrary finite sequence of disjoint bounded measurable sets in the space  $Y: B_Y^i B_Y^j = \emptyset$   $i, j = 1, \dots, s, i \neq j$ ,*

then

- 1\*. *the counting measure  $N^*(B_Y^j)$ ,  $j = 1, \dots, s$ , of the empirical ordered marked point process  $\mathcal{D} = (\mathcal{E}^*, \mathfrak{X}^*, P^*)$  is distributed by the Poisson law with parametric measure  $\lambda \mu(B_Y^j)$ :*

$$P^*\{N^*(B_Y^j) = k_j\} = \frac{[\lambda \mu(B_Y^j)]^{k_j}}{k_j!} e^{-\lambda \mu(B_Y^j)},$$

where  $\mu(B_Y^j) = P_y(B_Y^j)$ ,  $k_j = 0, 1, 2, \dots$ ;

- 2\*. *the counting measures  $\{N^*(B_Y^j), j = 1, \dots, s\}$  are jointly independent random variables:*

$$P^*\{N^*(B_Y^j) = k_j, j = 1, \dots, s\} = \prod_{j=1}^s P^*\{N^*(B_Y^j) = k_j\}.$$

*Proof.* The first statement is proved with the help of the full probability formula in view of the binomial distribution (34):

$$\begin{aligned} P^*\{N^*(B_Y^j) = k_j\} &= \sum_{n \geq k_j} P^*\{N^*(B_Y^j) = k_j \mid N^* = n\} P^*\{N^* = n\} \\ &= \sum_{n-k_j \geq 0} \frac{n!}{k_j! (n-k_j)!} \mu^{k_j}(B_Y^j) [1 - \mu(B_Y^j)]^{n-k_j} \frac{\lambda^{n-k_j} \lambda^{k_j}}{n!} e^{-\lambda} \\ &= \frac{[\lambda \mu(B_Y^j)]^{k_j}}{k_j!} e^{-\lambda} \sum_{m \geq 0} \frac{[\lambda(1 - \mu(B_Y^j))]^m}{m!}, \end{aligned}$$

where  $m = n - k_j$ . Since

$$\sum_{m \geq 0} \frac{[\lambda(1 - \mu(B_Y^j))]^m}{m!} = e^\lambda e^{-\lambda\mu(B_Y^j)},$$

we obtain

$$P^* \{N^*(B_Y^j) = k_j\} = \frac{[\lambda\mu(B_Y^j)]^{k_j}}{k_j!} e^{-\lambda\mu(B_Y^j)}.$$

To prove the second statement about the joint independence of the random variables  $\{N^*(B_Y^j), j = 1, \dots, s\}$  we use the full probability formula, polynomial law (35), and assume that  $B_Y^{s+1} = (\bigcup_{j=1}^s B_Y^j)^c$ ,  $k = \sum_{j=1}^s k_j$ ,  $k_{s+1} = n - k$  for  $n \geq k$ . Then

$$\begin{aligned} (43) \quad & P^* \{N^*(B_Y^j) = k_j, j = 1, \dots, s\} \\ &= \sum_{n \geq k} P^* \{N^*(B_Y^j) = k_j, j = 1, \dots, s+1 \mid N^* = n\} P^* \{N^* = n\} \\ &= \sum_{n \geq k} \frac{n!}{k_1! \dots k_s! (n-k)!} \mu^{k_1}(B_Y^1) \dots \mu^{k_s}(B_Y^s) \mu^{n-k}(B_Y^{s+1}) \frac{\lambda^k \lambda^{n-k}}{n!} e^{-\lambda} \\ &= \sum_{n-k \geq 0} \frac{n!}{k_1! \dots k_s! (n-k)!} \mu^{k_1}(B_Y^1) \dots \mu^{k_s}(B_Y^s) \mu^{n-k}(B_Y^{s+1}) \frac{\lambda^{k_1+\dots+k_s} \lambda^{n-k}}{n!} e^{-\lambda} \\ &= \prod_{j=1}^s \frac{[\lambda\mu(B_Y^j)]^{k_j}}{k_j!} e^{-\lambda} \left[ \sum_{m \geq 0} \frac{[\lambda\mu(B_Y^{s+1})]^m}{m!} \right] \quad (m = n - k). \end{aligned}$$

Since

$$\sum_{m \geq 0} \frac{[\lambda\mu(B_Y^{s+1})]^m}{m!} = e^{\lambda\mu(B_Y^{s+1})}, \quad \sum_{j=1}^{s+1} \mu(B_Y^j) = 1,$$

we conclude that

$$\begin{aligned} (44) \quad \exp\{-\lambda(\cdot)\} &= \exp\left\{-\lambda\left(\sum_{j=1}^{s+1} \mu(B_Y^j)\right)\right\} \exp\{\lambda\mu(B_Y^{s+1})\} \\ &= \exp\left\{-\lambda\sum_{j=1}^s \mu(B_Y^j)\right\}. \end{aligned}$$

Substituting (44) into (43) we get

$$\begin{aligned} P^* \{N^*(B_Y^j) = k_j, j = 1, \dots, s\} &= \prod_{j=1}^s \frac{[\lambda\mu(B_Y^j)]^{k_j}}{k_j!} e^{-\lambda\mu(B_Y^j)} \\ &= \prod_{j=1}^s P^* \{N^*(B_Y^j) = k_j\}. \quad \square \end{aligned}$$

**Definition 9.1.** A random process  $\mathcal{D} = (\mathcal{E}^*, \mathfrak{X}^*, P^*)$  satisfying the assumptions of Theorem 9.1 is called a strictly simple mixed empirical Poisson ordered marked point process with independent marks in a bounded space  $(Y, \mathfrak{A}_Y, \mathcal{B}_Y)$  [2, 3].

**Corollary 9.1.** *If the random variable  $N^*$  is distributed by the homogeneous Poisson law with parameter  $\lambda$ , then, using the general results (37)–(42), we obtain the moment*

characteristics of the mixed empirical Poisson ordered marked point process  $\mathcal{D}$  with independent marks:

$\nu_{\mathcal{D}}^{(1)}(B_Y^j) = \nu_{\mathcal{D}}^{(1)}(B_X^j \times B_K^j) = \lambda\mu(B_X^j \times B_K^j) = \lambda\mu_1(B_X^j)\mu_2(B_K^j) = \nu_{\mathcal{D}}^{(1)}(B_X^j)\mu_2(B_K^j)$ ,  
 where  $\nu_{\mathcal{D}}^{(1)}(B_X^j) = \lambda\mu_1(B_X^j)$  is the moment measure of the first order of the Poisson ordered point process  $\tilde{\mathcal{D}}$  of state points considered in Section 5 of [1],

$$\begin{aligned} \nu_{\mathcal{D}}^{(h)}(B_Y^{j_1} \times \cdots \times B_Y^{j_h}) &= \lambda^h \mu(B_Y^{j_1}) \cdots \mu(B_Y^{j_h}) = \nu_{\mathcal{D}}^{(1)}(B_Y^{j_1}) \cdots \nu_{\mathcal{D}}^{(1)}(B_Y^{j_h}), \\ \lambda_{\mathcal{D}}^{(h)}(B_Y^j) &= \lambda^h \mu^h(B_Y^j) = [\nu_{\mathcal{D}}^{(1)}(B_Y^j)]^h, \quad \nu_{\mathcal{D}}^{(2)}(B_Y^j) = \lambda^2 \mu^2(B_Y^j) + \lambda\mu(B_Y^j), \\ \nu_{\mathcal{D}}^{(2)}(B_Y^{j_1} \times B_Y^{j_2}) &= \lambda^2 \mu(B_Y^{j_1})\mu(B_Y^{j_2}), \\ D(N^*(B_Y^j)) &= \lambda\mu(B_Y^j), \quad \text{cov}[N^*(B_Y^{j_1}), N^*(B_Y^{j_2})] = 0. \end{aligned}$$

#### 10. MIXED EMPIRICAL NEGATIVE BINOMIAL ORDERED MARKED POINT PROCESS WITH INDEPENDENT MARKS

**Definition 10.1.** A random process  $\mathcal{D} = (\mathcal{E}^*, \mathfrak{X}^*, P^*)$  is called a strictly simple mixed empirical negative binomial ordered marked point process with independent marks in a bounded space  $(Y, \mathfrak{A}_Y, \mathcal{B}_Y)$  if the size of a sample  $N^*$  is a random variable with the negative binomial distribution.

Similarly to Section 6 of [1], one can evaluate the following probability generating functions:

$$(45) \quad \Pi_{N^*(B_Y^1), \dots, N^*(B_Y^s)}(z_1, \dots, z_s) = \left[ 1 + \beta \sum_{j=1}^s \mu(B_Y^j)(1 - z_j) \right]^{-\alpha},$$

$$(46) \quad \Pi_{N^*(B_Y^j)}(z_j) = \left[ 1 + \beta\mu(B_Y^j)(1 - z_j) \right]^{-\alpha}, \quad j = 1, \dots, s.$$

Then, using (37)–(42) and (45), (46), we obtain the following moment characteristics for a family of bounded measurable disjoint sets  $\{B_Y^j, j = 1, \dots, s\}$  that form a partition of the phase space  $Y = X \times K$ :

$$\begin{aligned} \nu_{\mathcal{D}}^{(h)}(B_Y^{j_1} \times \cdots \times B_Y^{j_h}) &= \beta^h \prod_{i=1}^h (\alpha + i - 1) \mu(B_Y^{j_1}) \cdots \mu(B_Y^{j_h}) \\ 1 \leq j_1 < \cdots < j_h \leq s, \quad h &= 1, \dots, s, \\ \alpha_{\mathcal{D}}^{(h)}(B_Y^j) &= \beta^h \prod_{i=1}^h (\alpha + i - 1) \mu^h(B_Y^j), \quad \nu_{\mathcal{D}}^{(1)}(B_Y^j) = \lambda\beta\mu(B_Y^j), \\ \nu_{\mathcal{D}}^{(2)}(B_Y^j) &= \lambda\beta\mu(B_Y^j)[(\alpha + 1)\beta\mu(B_Y^j) + 1], \\ \nu_{\mathcal{D}}^{(2)}(B_Y^{j_1} \times B_Y^{j_2}) &= \alpha(\alpha + 1)\beta^2\mu(B_Y^{j_1})\mu(B_Y^{j_2}), \\ D(N^*(B_Y^j)) &= \alpha\beta\mu(B_Y^j)[\beta\mu(B_Y^j) + 1], \\ (47) \quad \text{cov}[N^*(B_Y^{j_1}), N^*(B_Y^{j_2})] &= \alpha\beta^2\mu(B_Y^{j_1})\mu(B_Y^{j_2}), \quad 1 \leq j_1, j_2 \leq s, j_1 \neq j_2. \end{aligned}$$

Considering (45)–(47) we make the following conclusions:

- a) Counting measures  $N^*(B_Y^1), \dots, N^*(B_Y^s)$  form a family of mutually correlated identically distributed random variables with negative binomial distribution with parameters  $\beta\mu(B_Y^j) > 0, \alpha > 0, j = 1, \dots, s$ .

- b) There is a positive correlation between the counting measures  $N^*(B_Y^i)$  and  $N^*(B_Y^r)$  ( $1 \leq i, r \leq s, i \neq r$ ).

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