

A NOTE ON POISSON APPROXIMATION IN MULTIVARIATE CASE

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0. Introduction.

It has been studied in recent years to show the Poisson approximation for the sum of independent Bernoulli random variables which may or may not be identically distributed (see [1], [2], [4]).

In paper [3], K. Kawamura has derived sufficient conditions of a Poisson approximation for the sum of independent identically multivariate Bernoulli random variables. In this paper, we are going to extend the result of paper [2] and generalize the result of paper [3] to the multivariate case.

1. Notations and Definitions.

a. Suffix and n -dimensional vectors.

1. j, k, m, n : positive integers,
2. λ_i : parameter of Poisson distribution for every $i \in E$,
3. e_1, e_2, \dots, e_n : base of n -dimensional vectors,
4. $E = \{0, 1\}^n - \{O\}$ and $EO = \{0, 1\}^n$,
5. O : n -dimensional zero vector.
6. $i = (i_1, i_2, \dots, i_n)$: n -dimensional vector belonging to E ,
7. $k = (k_1, k_2, \dots, k_n)$: n -dimensional vector belonging to E ,
8. $s = (s_1, s_2, \dots, s_n)$: n -dimensional vector with nonnegative integer components.

b. Sum of Bernoulli vectors.

1. $\{X_{kj} = (X1_{kj}, X2_{kj}, \dots, Xn_{kj}), j=1, 2, \dots, n_k, k \geq 1\}$ be a sequence of independent multivariate Bernoulli vectors with

$$P_{kj}(i) = P(X_{kj} = i), \quad \text{for all } i \in EO,$$

where

$$\sum_{i \in EO} P_{kj}(i) = 1,$$

2. $P_j(i)$: $P_{kj}(i)$ expressed in the notation b.1. will be replaced by $P_j(i)$ for simplicity if we don't need any information about fixed k ,

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3. $S_k = \sum_{j=1}^{n_k} X_{kj}$: the sum of Bernoulli vectors,
4. n_k : positive integer with $\lim_{k \rightarrow \infty} n_k = \infty$.

c. Probability of the sum of Bernoulli vectors.

1. α_i : frequence of the observation i in n_k trials of $\{X_{kj}=(X1_{kj}, X2_{kj}, \dots, Xn_{kj}), j=1, 2, \dots, n_k, k \geq 1\}$,
2. $\alpha=(\alpha_{e_1}, \alpha_{e_2}, \dots, \alpha_i, \dots, \alpha_{e_1+\dots+e_n})$: 2^n-1 dimensional vector,
3. $i \cdot e_r$: inner product of i and e_r ,
4. $[C]=[\alpha; \sum_{i \cdot e_r=1} \alpha_i = s_r, r=1, 2, \dots, n, i \in E]$, where $[C]$ is a set of α uniquely defined by the given vector s ,
5. $f_t(i)$: trial number for the t -th occurrence of observation i in n_k trials with $f_t(i) \in \{1, 2, \dots, n_k\}$ and $t=1, 2, \dots, \alpha_i$,
6. $F_i = \{(f_1(i), f_2(i), \dots, f_{\alpha_i}(i)); 1 \leq f_1(i) < f_2(i) < \dots < f_{\alpha_i}(i) \leq n_k\}$
7. G_i : the set of integers expressed in $(f_1(i), f_2(i), \dots, f_{\alpha_i}(i))$ belonging to F_i , with $G_i = \{f_1(i), f_2(i), \dots, f_{\alpha_i}(i)\}$,
8. $T(i) = i_1 \cdot 2^0 + i_2 \cdot 2^1 + \dots + i_m \cdot 2^{m-1} + \dots + i_n \cdot 2^{n-1}$: one to one correspondence on E and S^n , where $S^n = \{1, 2, \dots, 2^n - 1\}$,
9. $i' < i \stackrel{def}{\iff} T(i') < T(i)$,
10. $H_i = \bigcup_{i' < i} G_{i'}$ where $i' < i$ is defined in the notation c.9.,
11. $Q_i(i) = P_{f_t(i)}(i)$ for simplicity,
12. $Q'_i(i) = P_{f_t(i)}(i) / P_{f_t(i)}(\mathbf{O})$ for simplicity,

$$13. P[S_k = s] = \sum_{[C]} \left\{ \sum_{F_{e_1}}^{\alpha_{e_1}} \prod_{t=1}^{\alpha_{e_1}} Q_t(e_1) \cdots \sum_{\substack{F_i \\ G_i \cap H_i = \emptyset}}^{\alpha_i} \prod_{t=1}^{\alpha_i} Q_t(i) \cdots \right. \\ \left. \sum_{\substack{F_{e_1+\dots+e_n} \\ G_{e_1+\dots+e_n} \cap H_{e_1+\dots+e_n} = \emptyset}}^{\alpha_{e_1+\dots+e_n}} \prod_{t=1}^{\alpha_{e_1+\dots+e_n}} Q_t(e_1+\dots+e_n) \right\} \prod_{\substack{j=1 \\ j \notin \cup G_i \\ i \in E}}^{n_k} P_j(\mathbf{O})$$

$P[S_k = s]$ will appear in section 2 in detail.

d. Variation forms of the probability $P[S_k = s]$

Let us express two variation forms of $P[S_k = s]$ for the proof of Poisson approximation.

1. Let us put

$$B_{n_k}(\alpha) = \sum_{F_{e_1}}^{\alpha_{e_1}} \prod_{t=1}^{\alpha_{e_1}} Q_t(e_1) \cdots \sum_{\substack{F_i \\ G_i \cap H_i = \emptyset}}^{\alpha_i} \prod_{t=1}^{\alpha_i} Q_t(i) \cdots \\ \sum_{\substack{F_{e_1+\dots+e_n} \\ G_{e_1+\dots+e_n} \cap H_{e_1+\dots+e_n} = \emptyset}}^{\alpha_{e_1+\dots+e_n}} \prod_{t=1}^{\alpha_{e_1+\dots+e_n}} Q_t(e_1+\dots+e_n),$$

then we have from notation c.13.

$$P[S_k = \mathbf{s}] = \sum_{[C]} \{B_{n_k}(\boldsymbol{\alpha})\} \prod_{\substack{j=1 \\ j \notin G_i \\ i \in E}}^{n_k} P_j(\mathbf{O})$$

2. Let us put

$$A_{n_k}(\boldsymbol{\alpha}) = \sum_{F_{e_1}} \prod_{l=1}^{\alpha_{e_1}} Q'_l(e_1) \cdots \sum_{F_i} \prod_{l=1}^{\alpha_i} Q'_l(i) \cdots \sum_{F_{e_1+\dots+e_n}} \prod_{l=1}^{\alpha_{e_1+\dots+e_n}} Q'_l(e_1+\dots+e_n),$$

$$G_{e_1+\dots+e_n} \cap H_{e_1+\dots+e_n} = \emptyset$$

then we have

$$P[S_k = \mathbf{s}] = \sum_{j=1}^{n_k} \{A_{n_k}(\boldsymbol{\alpha})\} \prod_{j=1}^{n_k} P_j(\mathbf{O}),$$

because $P[S_k = \mathbf{s}]$ (see notation c.13.) can be rewritten by

$$P[S_k = \mathbf{s}] = \sum_{[C]} \left\{ \sum_{F_{e_1}} \prod_{l=1}^{\alpha_{e_1}} Q'_l(e_1) \cdots \sum_{F_i} \prod_{l=1}^{\alpha_i} Q'_l(i) \cdots \sum_{F_{e_1+\dots+e_n}} \prod_{l=1}^{\alpha_{e_1+\dots+e_n}} Q'_l(e_1+\dots+e_n) \right\} \prod_{j=1}^{n_k} P_j(\mathbf{O}),$$

$$G_{e_1+\dots+e_n} \cap H_{e_1+\dots+e_n} = \emptyset$$

3. Let us denote $C_{n_k}(\boldsymbol{\alpha})$ from $B_{n_k}(\boldsymbol{\alpha})$ by removing all the restriction in the sums (see notation d.1.) as follows,

$$C_{n_k}(\boldsymbol{\alpha}) = \sum_{F_{e_1}} \prod_{l=1}^{\alpha_{e_1}} Q_l(e_1) \cdots \sum_{F_i} \prod_{l=1}^{\alpha_i} Q_l(i) \cdots \sum_{F_{e_1+\dots+e_n}} \prod_{l=1}^{\alpha_{e_1+\dots+e_n}} Q_l(e_1+\dots+e_n).$$

2. Conditions sufficient for Poisson approximation.

Let $\{X_{kj} = (X1_{kj}, X2_{kj}, \dots, Xn_{kj}), j=1, 2, \dots, n_k, k \geq 1\}$ be a sequence of independent multivariate Bernoulli vectors with

$$P_{kj}(i) = P(X_{kj} = i), \quad \text{for all } i \in \mathbf{EO},$$

where

$$\sum_{i \in \mathbf{EO}} P_{kj}(i) = 1,$$

and denote the sum of multivariate Bernoulli vectors by $S_k = \sum_{j=1}^{n_k} X_{kj}$. In the following discussion, $P_{kj}(i)$ expressed in notation b.1. will be replaced by $P_j(i)$ for simplicity and we can easily see that

$$P[S_k = \mathbf{s}] = \sum_{[C]} \left\{ \sum_{F_{e_1}} \prod_{t=1}^{\alpha_{e_1}} Q_t(e_1) \cdots \sum_{F_{e_i}} \prod_{t=1}^{\alpha_i} Q_t(i) \cdots \right. \\ \left. \sum_{\substack{F_{e_1+\dots+e_n} \\ G_{e_1+\dots+e_n} \cap H_{e_1+\dots+e_n} = \emptyset}} \prod_{t=1}^{\alpha_{e_1+\dots+e_n}} Q_t(e_1+\dots+e_n) \right\} \prod_{\substack{j=1 \\ j \notin \cup G_i \\ i \in E}}^{n_k} P_j(\mathbf{O})$$

THEOREM. *If the following relations (2.1) and (2.2) are satisfied for the sequence of independent n -variate Bernoulli distribution which may or may not be identically distributed, that is, for every $\mathbf{i} \in \mathbf{E}$*

$$(2.1) \quad \sum_{j=1}^{n_k} P_{kj}(\mathbf{i}) \rightarrow \lambda_{\mathbf{i}} \quad \text{as } k \rightarrow \infty,$$

$$(2.2) \quad \min_{1 \leq j \leq n_k} P_{kj}(\mathbf{i}) \rightarrow 1 \quad \text{as } k \rightarrow \infty,$$

then we have

$$(2.3) \quad \lim_{k \rightarrow \infty} P[S_k = \mathbf{s}] = \sum_{[C]} \left[\prod_{\mathbf{i} \in \mathbf{E}} (\lambda_{\mathbf{i}}^{\alpha_{\mathbf{i}}} / \alpha_{\mathbf{i}}!) \right] \exp\left(-\sum_{\mathbf{i} \in \mathbf{E}} \lambda_{\mathbf{i}}\right)$$

for all \mathbf{s} ($\mathbf{s} \geq 0$), where $[C]$ is uniquely defined by the vector \mathbf{s} as

$$[C] = [\boldsymbol{\alpha}; \sum_{\mathbf{i} \cdot \mathbf{e}_r = 1} \alpha_{\mathbf{i}} = s_r, r = 1, 2, \dots, n, \mathbf{i} \in \mathbf{E}].$$

In order to prove the theorem, we are going to show lemma 1, lemma 2 and lemma 3.

LEMMA 1. *If the conditions (2.1) and (2.2) are satisfied then we have*

$$\sum_{g=1}^{n_k} P_g(\mathbf{O}) \rightarrow \exp\left(-\sum_{\mathbf{i} \in \mathbf{E}} \lambda_{\mathbf{i}}\right) \quad \text{as } k \rightarrow \infty.$$

Proof. Consider the inequality

$$1 + y \leq \exp(y), \quad y \in [-1, \infty),$$

putting $y = -x$ and $y = x/(1-x)$ with $x \in [0, 1]$ we obtain

$$\exp(-x/(1-x)) \leq 1 - x \leq \exp(-x), \quad x \in [0, 1].$$

Now putting $\Delta_g = \sum_{\mathbf{i} \in \mathbf{E}} P_g(\mathbf{i}) = 1 - P_g(\mathbf{O})$,

where $0 \leq \Delta_g < 1$ (by (2.2)) for sufficiently large k ($1 \leq g \leq n_k$), and using the last inequality, we get

$$\exp\left[-\left(\sum_{g=1}^{n_k} \Delta_g\right) / \min_g P_g(\mathbf{O})\right] \leq \prod_{g=1}^{n_k} P_g(\mathbf{O}) \leq \exp\left(-\sum_{g=1}^{n_k} \Delta_g\right),$$

and from (2.1), (2.2) we can prove that

$$(2.4) \quad \sum_{g=1}^{n_k} P_g(\mathbf{O}) \rightarrow \exp(-\sum_{i \in E} \lambda_i) \quad \text{as } k \rightarrow \infty. \quad \blacksquare$$

LEMMA 2. *If the conditions (2.1) and (2.2) are satisfied then we have*

$$(2.5) \quad C_{n_k}(\boldsymbol{\alpha}) \rightarrow \prod_{i \in E} (\lambda_i^{\alpha_i} / \alpha_i!) \quad \text{as } k \rightarrow \infty.$$

Proof. It is sufficient to prove that

$$(2.5.1) \quad \sum_{F_i} [\prod_{t=1}^{\alpha_i} P_{f_t(i)}(\mathbf{i})] \rightarrow \lambda_i^{\alpha_i} / \alpha_i!, \quad \text{for every } \mathbf{i} \in E.$$

The proof of (2.5.1) is given by induction with respect to α_i .

(1) $\alpha_i=1$. It is obvious by (2.1) that

$$\sum_{f_1(i)=1}^{n_k} P_{f_1(i)}(\mathbf{i}) \rightarrow \lambda_i \quad \text{as } k \rightarrow \infty.$$

(2) $\alpha_i=2$. By (2.1) and (2.2), we have

$$\sum_{f_1(i) < f_2(i)} P_{f_1(i)}(\mathbf{i}) P_{f_2(i)}(\mathbf{i}) \rightarrow (\lambda_i)^2 / 2!$$

because

$$0 \leq \sum_{j=1}^{n_k} P_j^2(\mathbf{i}) \leq [1 - \min_j P_j(\mathbf{O})] \sum_{j=1}^{n_k} P_j(\mathbf{i})$$

and by (2.1), (2.2) the right hand side of the inequality tends to 0, so we have

$$2 \sum_{f_1(i) < f_2(i)} P_{f_1(i)}(\mathbf{i}) P_{f_2(i)}(\mathbf{i}) = [\sum_{j=1}^{n_k} P_j(\mathbf{i})]^2 - \sum_{j=1}^{n_k} P_j^2(\mathbf{i}) \rightarrow \lambda_i^2 \quad \text{as } k \rightarrow \infty.$$

(3) Assume that (2.5.1) is correct as $\alpha_i=m-1$, that is,

$$\sum_{f_1(i) < \dots < f_{m-1}(i)} \prod_{t=1}^{m-1} P_{f_t(i)}(\mathbf{i}) \rightarrow (\lambda_i)^{m-1} / (m-1)! \quad \text{as } k \rightarrow \infty.$$

In order to finish the induction, let us prove (2.5.1) to be also true as $\alpha_i=m$.

Multiply the left hand side of the last relation by $\sum_{f_m(i)=1}^{n_k} P_{f_m(i)}(\mathbf{i})$ which tends to λ_i (by (2.1)), we obtain

$$\begin{aligned} & \sum_{f_1(i) < \dots < f_{m-1}(i)} P_{f_1(i)}(\mathbf{i}) \prod_{t=1}^{m-1} P_{f_t(i)}(\mathbf{i}) \\ & + \sum_{f_1(i) < \dots < f_{m-1}(i)} P_{f_2(i)}(\mathbf{i}) \prod_{t=1}^{m-1} P_{f_t(i)}(\mathbf{i}) + \dots \\ (2.5.2) \quad & + \sum_{f_1(i) < \dots < f_{m-1}(i)} P_{f_{m-1}(i)}(\mathbf{i}) \prod_{t=1}^{m-1} P_{f_t(i)}(\mathbf{i}) \\ & + \sum_{f_m(i) < f_1(i) < \dots < f_{m-1}(i)} \prod_{t=1}^m P_{f_t(i)}(\mathbf{i}) \end{aligned}$$

$$\begin{aligned}
 &+ \sum_{f_1(i) < f_m(i) < f_2(i) < \dots < f_{m-1}(i)} \prod_{t=1}^m P_{f_t(i)}(\mathbf{i}) + \dots \\
 &+ \sum_{f_1(i) < \dots < f_{m-1}(i) < f_m(i)} \prod_{t=1}^m P_{f_t(i)}(\mathbf{i})
 \end{aligned}$$

Each of the first $(m-1)$ terms of (2.5.2) may be nonnegative and estimated by

$$[1 - \min P_j(\mathbf{O})] \sum_{f_1(i) < \dots < f_{m-1}(i)} \prod_{t=1}^{m-1} P_{f_t(i)}(\mathbf{i})$$

which is an upper bound of these terms and by (2.2) tends to 0, that is,

$$0 \leq [\text{each of the first } (m-1) \text{ terms of (2.5.2)}]$$

$$\leq [1 - \min P_j(\mathbf{O})] \sum_{f_1(i) < \dots < f_{m-1}(i)} \prod_{t=1}^{m-1} P_{f_t(i)}(\mathbf{i})$$

So each of the first $(m-1)$ terms tends to 0, and each of the last m terms has the same value, then we can obtain the limiting value of (2.5.2) to be

$$m \sum_{f_1(i) < \dots < f_{m-1}(i)} \prod_{j=1}^m P_{f_j(i)}(\mathbf{i}) \rightarrow \lambda_i (\lambda_i)^{m-1} / (m-1)!,$$

that is, (2.5.1) is correct as $\alpha_i = m$ and we finish the proof of (2.5.1) by induction. Then by (2.5.1), we have

$$(2.5) \quad C_{n_k}(\boldsymbol{\alpha}) \rightarrow \prod_{i \in E} (\lambda_i^{\alpha_i} / \alpha_i!) \quad \text{as } k \rightarrow \infty.$$

this is the result of lemma 2. ■

LEMMA 3. Three values (defined in the notation d .) $A_{n_k}(\boldsymbol{\alpha})$, $B_{n_k}(\boldsymbol{\alpha})$ and $C_{n_k}(\boldsymbol{\alpha})$ have the same limiting value, that is,

$$(2.8) \quad B_{n_k}(\boldsymbol{\alpha}) \rightarrow \prod_{i \in E} (\lambda_i^{\alpha_i} / \alpha_i!) \quad \text{as } k \rightarrow \infty,$$

and

$$(2.9) \quad A_{n_k}(\boldsymbol{\alpha}) \rightarrow \prod_{i \in E} (\lambda_i^{\alpha_i} / \alpha_i!) \quad \text{as } k \rightarrow \infty.$$

Proof. For the proof of lemma 3, we consider the following three steps.
(step 1) Let us define

$$\begin{aligned}
 \text{Rem}(\mathbf{i}) &= \sum_{F_i} \prod_{t=1}^{\alpha_i} P_{f_t(i)}(\mathbf{i}) - \sum_{G_i \cap H_i = \emptyset} \prod_{t=1}^{\alpha_i} P_{f_t(i)}(\mathbf{i}), \\
 &= \sum_{F_i} \prod_{t=1}^{\alpha_i} P_{f_t(i)}(\mathbf{i}), \\
 &\quad G_i \cap H_i \neq \emptyset
 \end{aligned}$$

for every $\mathbf{i} \in E - \{e_1\}$. In this step, we are going to prove that

$$(2.6) \text{ Rem}(\mathbf{i}) \rightarrow 0 \text{ as } k \rightarrow \infty,$$

for every $\mathbf{i} \in \mathbf{E} - \{\mathbf{e}_1\}$.

Proof of (2.6): It is easy to see that

$$(2.7) \sum_{\mathbf{F}_i} \prod_{t=1}^n P_{f_t(\mathbf{i})}(\mathbf{i}) \leq \left[\sum_{f_t(\mathbf{i})=1}^{n_k} P_{f_t(\mathbf{i})}(\mathbf{i}) \right]^n \leq (\lambda_i + \varepsilon)^n.$$

It is obvious that $\text{Rem}(\mathbf{i})$ is nonnegative and estimated as follows:

$$\begin{aligned} \text{Rem}(\mathbf{i}) &\leq \sum_{r=1}^{d(\mathbf{i})} \sum_{s=1}^{\alpha_i} P_{f_s(\mathbf{i})}(\mathbf{i}) \sum_{\substack{\mathbf{F}_i \\ t=1 \\ \neq s}}^{\alpha_i} P_{f_t(\mathbf{i})}(\mathbf{i}) \\ &\leq d(\mathbf{i}) \alpha_i [1 - \min_j P_j(\mathbf{O})] (\lambda_i + \varepsilon)^{\alpha_i - 1} \end{aligned}$$

with $f_s(\mathbf{i}) = f_r(k)$, where $d(\mathbf{i}) = \sum_{i < k} \alpha_k$.

By (2.2) and (2.7) for $n = \alpha_i - 1$ the right hand side of the last inequality tends to zero as $k \rightarrow \infty$, and we finish step 1.

(step 2) By the definition of $C_{n_k}(\boldsymbol{\alpha})$ we can obtain

$$\begin{aligned} C_{n_k}(\boldsymbol{\alpha}) &= \prod_{\mathbf{i} \in \mathbf{E}} \sum_{\mathbf{F}_i} \prod_{t=1}^{\alpha_i} P_{f_t(\mathbf{i})}(\mathbf{i}) \\ &= \sum_{\mathbf{F}_{\mathbf{e}_1}} \prod_{t=1}^{\alpha_{\mathbf{e}_1}} P_{f_t(\mathbf{e}_1)}(\mathbf{e}_1) \left\{ \prod_{\mathbf{i} \in \mathbf{E} - \{\mathbf{e}_1\}} [\text{Rem}(\mathbf{i}) + \sum_{\substack{\mathbf{F}_i \\ t=1 \\ G_i \cap H_i = \emptyset}} \prod_{t=1}^{\alpha_i} P_{f_t(\mathbf{i})}(\mathbf{i})] \right\} \end{aligned}$$

and by the definitions of $\text{Rem}(\mathbf{i})$, $B_{n_k}(\boldsymbol{\alpha})$ and using (2.6), we can obtain that $B_{n_k}(\boldsymbol{\alpha})$ and $C_{n_k}(\boldsymbol{\alpha})$ have the same limiting value as $k \rightarrow \infty$. Then from lemma 2, we have

$$(2.8) \quad B_{n_k}(\boldsymbol{\alpha}) \rightarrow \prod_{\mathbf{i} \in \mathbf{E}} (\lambda_i^{\alpha_i} / \alpha_i!) \text{ as } k \rightarrow \infty.$$

(step 3) It is easy to see by the definition of $A_{n_k}(\boldsymbol{\alpha})$, $B_{n_k}(\boldsymbol{\alpha})$ that

$$B_{n_k}(\boldsymbol{\alpha}) \leq A_{n_k}(\boldsymbol{\alpha}) \leq (1 / \min_j P_j(\mathbf{O}))^h B_{n_k}(\boldsymbol{\alpha}),$$

where $h = \sum_{\mathbf{i} \in \mathbf{E}} \alpha_i$ and by (2.2), (2.8), we have

$$(2.9) \quad A_{n_k}(\boldsymbol{\alpha}) \rightarrow \prod_{\mathbf{i} \in \mathbf{E}} (\lambda_i^{\alpha_i} / \alpha_i!) \text{ as } k \rightarrow \infty. \quad \blacksquare$$

Proof of the theorem.

Summarize lemma 1 and lemma 3, we finish the theorem. \blacksquare

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