## CORRECTION

## COMPOUND POISSON APPROXIMATIONS FOR SUMS OF RANDOM VARIABLES

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The proof of Theorem 1.1 contains the erroneous statement that

$$P\bigg(\sum_{i=1}^n X_i \in B\bigg) = E\big[F_1' * \cdots * F_n'(B)\big].$$

The following modifications to Theorem 1.1 and Lemma 2.2 will correct this error. To simplify the notation, we let  $d(X|\mathcal{F}, Y|\mathcal{G})$  denote the random total-variation distance  $d(P(X \in \mathcal{F}), P(Y \in \mathcal{G}))$ .

(a) For the statement of Theorem 1.1, let  $\zeta_i = I(X_i \neq 0)$  and redefine  $F_i$  as  $F_i(B) = P(X_i \in B | \mathcal{F}_{i-1}, X_i \neq 0, \zeta_h, k \neq i)$ .

Then in the bounds (1.3), (1.4), (1.5), replace each  $d_i$  by  $\zeta_i d_i$ . The error was that the original  $F_i$  did not contain the additional conditioning on  $\zeta_k$ ,  $k \neq i$ , which is needed in the proof.

- (b) Replace assertion (iv) of Lemma 2.2 by the following:
- (iv) If  $X_1, \ldots, X_n$  are adapted to the increasing  $\sigma$ -fields  $\{\mathscr{F}_i\}_{i=0}^n$  and  $Y_1, \ldots, Y_n$  are independent with respective distributions  $G_1, \ldots, G_n$ , then

(2.4) 
$$D \leq E\left[\sum_{i=1}^{n} d(X_{i}|\mathscr{F}_{i-1}, Y_{i})\right].$$

(There is no conditioning on the  $Y_{i}$ .)

(c) Replace the last paragraph of the proof of Lemma 2.2 by the following:

We prove statement (iv) by induction. It is true for n=1 since  $d(X_1,Y_1) \le E[d(X_1|\mathscr{F}_0,Y_1)]$  by Lemma 2.1. Now assume it is true for n-1. Let  $X^*=(X_1,\ldots,X_{n-1},Y_n)$ , where  $Y_n$  is independent of  $X_1,\ldots,X_{n-1}$ . Then by (i) and Lemma 2.1, we have

$$\begin{split} D &\leq d(X,Y) \leq d(X,X^*) + d(X^*,Y) \leq E\left[d(X|\mathscr{F}_{n-1},X^*|\mathscr{F}_{n-1})\right] + d(X^*,Y) \\ &= E\left[d(X_n|\mathscr{F}_{n-1},Y_n)\right] + d((X_1,\ldots,X_{n-1}),(Y_1,\ldots,Y_{n-1})), \end{split}$$

which equals the right-hand side of (2.4).

(d) In the proof of Theorem 1.1, replace the five lines after expression (3.1) with the following:

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We can write  $Y = \sum_{i=1}^n \zeta_i Y_i$ , where  $Y_1, \ldots, Y_n$  are independent with the common distribution F and are independent of  $\zeta_1, \ldots, \zeta_n$ . Let  $\mathscr H$  denote the  $\sigma$ -field generated by  $\zeta_1, \ldots, \zeta_n$ . Then applying (2.2) and Lemma (2.2)(iv) with  $X_i, Y_i, P(\cdot)$  equal to  $\zeta_i X_i, \zeta_i Y_i, P(\cdot|\mathscr H)$ , respectively, we have

$$d\left(\sum_{i=1}^{n} X_{i}, Y\right) = d\left(\sum_{i=1}^{n} \zeta_{i} X_{i}, \sum_{i=1}^{n} \zeta_{i} Y_{i}\right)$$

$$\leq E\left[d\left(\sum_{i=1}^{n} \zeta_{i} X_{i} | \mathcal{H}, \sum_{i=1}^{n} \zeta_{i} Y_{i} | \mathcal{H}\right)\right]$$

$$\leq E\left\{\sum_{i=1}^{n} E\left[d\left(\zeta_{i} X_{i} | (\mathcal{F}_{i-1}, \mathcal{H}), \zeta_{i} Y_{i} | \mathcal{H}\right) | \mathcal{H}\right]\right\}$$

$$= E\left(\sum_{i=1}^{n} \zeta_{i} d_{i}\right).$$

To obtain the last inequality, first note that the standard properties of conditional probabilities allow us to write

$$P(\zeta_i X_i \in B | \mathscr{F}_{i-1}, \mathscr{H})$$

$$= P(X_i \in B | \mathscr{F}_{i-1}, \zeta_i = 1, \zeta_k, k \neq i) I(\zeta_i = 1) + I(0 \in B) I(\zeta_i = 0).$$

Then using this and a similar expression for  $P(\zeta_i Y_i \in B | \mathcal{H})$ , it follows from the definition of d that

$$d(\zeta_i X_i | (\mathscr{F}_{i-1}, \mathscr{H}), \zeta_i Y_i | \mathscr{H}) = I(\zeta_i = 1) d_i = \zeta_i d_i.$$

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