

# Instant-by-instant and simultaneous confidence prediction bands

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## Appendix

**Definition 1.** Let  $K$  the size of the fleet and  $T$  the size of the time grid be non-zero integers. Let a cumulative event number evolution curve  $\xi = (\xi_1, \dots, \xi_T)$  be a random vector where for every  $\tau \in \{1, \dots, T\}$ ,  $\xi_\tau$  is a  $\{1, \dots, K\}$ -valued random variable. Let a band  $B = (B_1, \dots, B_T)$  be such that for every  $\tau \in \{1, \dots, T\}$ ,  $\hat{B}_{\alpha;\tau} \subseteq \{1, \dots, K\}$ .

**Definition 2** (Instant-by-instant coverage). Let  $\alpha \in (0, 1)$ . A band  $B$  controls the instant-by-instant coverage of the curve  $\xi$  at a control level  $\alpha$  if

$$\forall \tau \in \{1, \dots, T\}, \quad \mathbb{P} \left[ \xi_\tau \in \hat{B}_{\alpha;\tau} \right] \geq 1 - \alpha.$$

**Definition 3** ( $\gamma$ -Simultaneous coverage). Let  $\alpha \in (0, 1)$  and  $\gamma \in [0, 1]$ . A band  $B$  controls the  $\gamma$ -simultaneous coverage of the curve  $\xi$  at a control level  $\alpha$  if

$$\mathbb{P} \left[ \frac{\text{Card} \left( \left\{ \tau \in \{1, \dots, T\} : \xi_\tau \in \hat{B}_{\alpha;\tau} \right\} \right)}{T} \geq 1 - \gamma \right] \geq 1 - \alpha.$$

**Proposition 1** (Instant-by-instant prediction band, MD-full). Let  $\xi_1, \dots, \xi_n$  be  $n$  independent copies of  $\xi$ . For any control level  $\alpha \in [0, 1]$ , the band  $\hat{B}_{\alpha}^{\text{MD-full}}$  defined as, for every  $\tau \in \{1, \dots, T\}$

$$\hat{B}_{\alpha;\tau}^{\text{MD-full}} := \left\{ k \in \{1, \dots, K\} : \hat{p}_\tau(k) \geq \hat{\ell}_{\alpha;\tau} - \frac{1}{n} \right\} \quad (1)$$

ensures control of the instant-by-instant coverage of  $\xi$  at a control level  $\alpha$ , where the marginal empirical probability density function  $\hat{p}_\tau$  is defined in Equation (Eq. (2)) and the threshold  $\hat{\ell}_{\alpha;\tau}$  in Equation (Eq. (3)).

*Proof of Proposition 1.* Let  $\alpha \in [0, 1]$  and  $\tau \in \{1, \dots, T\}$ . Following Lei, Robins, and Wasserman (2013), a full conformal prediction  $\hat{C}_{\alpha;\tau}$  region is built from the observations  $\xi_{1;\tau}, \dots, \xi_{n;\tau}$  for  $\xi_\tau$  as follows

$$\hat{C}_{\alpha;\tau} = \{k \in \{1, \dots, K\} : \hat{\pi}_\tau(k) > \alpha\}$$

where for every  $k \in \{1, \dots, K\}$

$$\hat{\pi}_\tau(k) := \frac{1 + \sum_{i=1}^n \mathbb{1} \{ \hat{p}_\tau^k(\xi_{i;\tau}) \leq \hat{p}_\tau^k(k) \}}{n + 1}$$

where for every  $l \in \{1, \dots, K\}$

$$\hat{p}_\tau^k(l) = \frac{n}{n+1} \hat{p}_\tau(l) + \frac{1}{n+1} \mathbb{1}\{k=l\},$$

and the empirical marginal probability density function is given by

$$\hat{p}_\tau(l) = \frac{1}{n} \sum_{i=1}^n \mathbb{1}\{\xi_{i;\tau} = l\} \quad (2)$$

Since  $\xi_{1;\tau}, \dots, \xi_{n;\tau}$  are i.i.d. copies of  $\xi_\tau$ ,  $\hat{C}_{\alpha;\tau}$  is a confidence prediction region (Vovk, Gamerman, and Shafer, 2005)

$$\mathbb{P}\left[\xi_\tau \in \hat{C}_{\alpha;\tau}\right] \geq 1 - \alpha.$$

Let us now provide an explicit expression of this region.

$$\begin{aligned} k \in \hat{C}_{\alpha;\tau} &\iff \hat{\pi}_\tau(k) > \alpha \\ &\iff \frac{1 + \sum_{i=1}^n \mathbb{1}\{\hat{p}_\tau^k(\xi_{i;\tau}) \leq \hat{p}_\tau^k(k)\}}{n+1} > \alpha \\ &\iff \frac{1}{n} \sum_{i=1}^n \mathbb{1}\{\hat{p}_\tau^k(\xi_{i;\tau}) > \hat{p}_\tau^k(k)\} < (1-\alpha) \left(1 + \frac{1}{n}\right). \end{aligned}$$

Let us consider the term on the l.h.s.

$$\begin{aligned} &\frac{1}{n} \sum_{i=1}^n \mathbb{1}\{\hat{p}_\tau^k(\xi_{i;\tau}) > \hat{p}_\tau^k(k)\} \\ &= \frac{1}{n} \sum_{i=1}^n \sum_{l=1}^K \mathbb{1}\{l = \xi_{i;\tau}\} \mathbb{1}\{\hat{p}_\tau^k(\xi_{i;\tau}) > \hat{p}_\tau^k(k)\} \\ &= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1}\{\hat{p}_\tau^k(l) > \hat{p}_\tau^k(k)\} \\ &= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1}\left\{\frac{n}{n+1} \hat{p}_\tau(l) + \frac{1}{n+1} \mathbb{1}\{k=l\} > \frac{n}{n+1} \hat{p}_\tau(k) + \frac{1}{n+1}\right\} \\ &= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1}\left\{\hat{p}_\tau(l) + \frac{1}{n} \mathbb{1}\{k=l\} > \hat{p}_\tau(k) + \frac{1}{n}\right\} \\ &= \hat{p}_\tau(k) \mathbb{1}\left\{\hat{p}_\tau(k) + \frac{1}{n} > \hat{p}_\tau(k) + \frac{1}{n}\right\} + \sum_{l=1, l \neq k}^K \hat{p}_\tau(l) \mathbb{1}\left\{\hat{p}_\tau(l) > \hat{p}_\tau(k) + \frac{1}{n}\right\} \\ &= \hat{p}_\tau(k) \mathbb{1}\left\{\hat{p}_\tau(k) > \hat{p}_\tau(k) + \frac{1}{n}\right\} + \sum_{l=1, l \neq k}^K \hat{p}_\tau(l) \mathbb{1}\left\{\hat{p}_\tau(l) > \hat{p}_\tau(k) + \frac{1}{n}\right\} \\ &= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1}\left\{\hat{p}_\tau(l) > \hat{p}_\tau(k) + \frac{1}{n}\right\}. \end{aligned}$$

It follows that

$$k \in \hat{C}_{\alpha;\tau} \iff \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1}\left\{\hat{p}_\tau(l) > \hat{p}_\tau(k) + \frac{1}{n}\right\} < (1-\alpha) \left(1 + \frac{1}{n}\right)$$

$$\iff \hat{p}_\tau(k) \geq \hat{\ell}_{\alpha;\tau},$$

where the threshold  $\hat{\ell}_{\alpha;\tau}$  is defined as

$$\hat{\ell}_{\alpha;\tau} := \inf \left\{ t \in [0, 1] : \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1} \{ \hat{p}_\tau(l) > t \} < (1 - \alpha) \left( 1 + \frac{1}{n} \right) \right\}. \quad (3)$$

**How to compute  $\hat{\ell}_{\alpha;\tau}$ .** The function  $G : [0, 1] \rightarrow [0, 1]$ ,  $t \mapsto \hat{G}_\tau(t) := \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1} \{ \hat{p}_\tau(l) > t \}$  is decreasing piece-wise constant right-continuous function which changes value at  $\hat{p}_\tau(1), \dots, \hat{p}_\tau(K)$ . Let  $x_1 > \dots > x_\kappa$  be the distinct values among  $\hat{p}_\tau(1), \dots, \hat{p}_\tau(K)$  sorted from largest to smallest, and for all  $j \in \{1, \dots, \kappa\}$ ,  $\rho_j := \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1} \{ \hat{p}_\tau(l) = x_j \}$ . Consider the cumulative sum of  $\rho_1, \dots, \rho_\kappa$ ,  $s_1, \dots, s_\kappa$ . It follows that  $G(x_1) = 0$  and for every  $j \in \{2, \dots, \kappa\}$

$$\begin{aligned} G(x_j) &= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1} \{ \hat{p}_\tau(l) > x_j \} = \sum_{l=1}^K \sum_{i=1}^{\kappa} \mathbb{1} \{ x_i = \hat{p}_\tau(l) \} \hat{p}_\tau(l) \mathbb{1} \{ \hat{p}_\tau(l) > x_j \} \\ &= \sum_{i=1}^{\kappa} \rho_i \mathbb{1} \{ x_i > x_j \} = \sum_{i=1}^{j-1} \rho_i = s_{j-1}. \end{aligned}$$

It follows that, one only has to check for finite number of values

$$\begin{aligned} \hat{\ell}_{\alpha;\tau} &= \inf \left\{ t \in [0, 1] : \hat{G}_\tau(t) < (1 - \alpha) \left( 1 + \frac{1}{n} \right) \right\} \\ &= \min \left\{ x_i : i \in \{1, \dots, \kappa\}, s_{i-1} < (1 - \alpha) \left( 1 + \frac{1}{n} \right) \right\}. \end{aligned}$$

□

**Proposition 2** (Instant-by-instant prediction upper band, MDist-full). *Let  $\xi_1, \dots, \xi_n$  be  $n$  independent copies of  $\xi$ . For any control level  $\alpha \in [0, 1]$ , the band  $\hat{B}_\alpha^{\text{MDist-full,up}}$  defined as, for every  $\tau \in \{1, \dots, T\}$*

$$\hat{B}_{\alpha;\tau}^{\text{MDist-full,up}} := \left\{ k \in \{1, \dots, K\} : k \leq \hat{Q}_\tau^{\text{up}} \left( (1 - \alpha) \left( 1 + \frac{1}{n} \right) \right) + 1 \right\} \quad (4)$$

*ensures control of the instant-by-instant coverage of  $\xi$  at a control level  $\alpha$ , the empirical marginal upper quantile function  $\hat{Q}_\tau^{\text{up}}$  is defined in Equation (Eq. (6)).*

*Proof of Proposition 2.* Let  $\alpha \in [0, 1]$  and  $\tau \in \{1, \dots, T\}$ . A full conformal prediction region  $\hat{C}_{\alpha;\tau}$  can be defined as

$$\hat{C}_{\alpha;\tau} := \{ k \in \{1, \dots, K\} : \hat{\pi}(k) > \alpha \},$$

where for every  $k \in \{1, \dots, K\}$

$$\hat{\pi}(k) := \frac{1 + \sum_{i=1}^n \mathbb{1} \left\{ \hat{F}_\tau^k(\xi_{i;\tau}) \geq \hat{F}_\tau^k(k) \right\}}{n + 1},$$

and for every  $l \in \{1, \dots, K\}$

$$\hat{F}_\tau^k(l) := \frac{n}{n+1} \hat{F}_\tau(l) + \frac{1}{n+1} \mathbb{1} \{ k \leq l \},$$

where the empirical marginal cumulative density function  $\hat{F}_\tau$  is defined as

$$\hat{F}_\tau(l) := \frac{1}{n} \sum_{i=1}^n \mathbb{1} \{ \xi_{i;\tau} \leq l \}. \quad (5)$$

Since  $\xi_{1;\tau}, \dots, \xi_{n;\tau}$  are i.i.d. copies of  $\xi_\tau$ ,  $\hat{C}_{\alpha;\tau}$  is a confidence prediction region (Vovk, Gamerman, and Shafer, 2005)

$$\mathbb{P} \left[ \xi_\tau \in \hat{C}_{\alpha;\tau} \right] \geq 1 - \alpha.$$

Let us now provide an explicit expression of  $\hat{C}_{\alpha;\tau}$ .

$$\begin{aligned} k \in \hat{C}_{\alpha;\tau} &\iff \hat{\pi}_\tau(k) > \alpha \\ &\iff \frac{1 + \sum_{i=1}^n \mathbb{1} \left\{ \hat{F}_\tau^k(\xi_{i;\tau}) \geq \hat{F}_\tau^k(k) \right\}}{n+1} > \alpha \\ &\iff \frac{1}{n} \sum_{i=1}^n \mathbb{1} \left\{ \hat{F}_\tau^k(\xi_{i;\tau}) < \hat{F}_\tau^k(k) \right\} < (1-\alpha) \left( 1 + \frac{1}{n} \right). \end{aligned}$$

Let us focus on the l.h.s. term. Let us define the function  $\hat{G}_\tau : \{1, \dots, K\} \rightarrow [0, 1]$  such that for every  $k$

$$\begin{aligned} \hat{G}_\tau(k) &:= \frac{1}{n} \sum_{i=1}^n \mathbb{1} \left\{ \hat{F}_\tau^k(\xi_{i;\tau}) < \hat{F}_\tau^k(k) \right\} \\ &= \frac{1}{n} \sum_{i=1}^n \sum_{l=1}^K \mathbb{1} \{ l = \xi_{i;\tau} \} \mathbb{1} \left\{ \hat{F}_\tau^k(\xi_{i;\tau}) < \hat{F}_\tau^k(k) \right\} \\ &= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1} \left\{ \hat{F}_\tau^k(l) < \hat{F}_\tau^k(k) \right\} \\ &= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1} \left\{ \frac{n}{n+1} \hat{F}_\tau(l) + \frac{1}{n+1} \mathbb{1} \{ k \leq l \} < \frac{n}{n+1} \hat{F}_\tau(k) + \frac{1}{n+1} \right\} \\ &= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1} \left\{ \hat{F}_\tau(l) + \frac{1}{n} \mathbb{1} \{ k \leq l \} < \hat{F}_\tau(k) + \frac{1}{n} \right\} \\ &= \sum_{l=1, l < k}^K \hat{p}_\tau(l) \mathbb{1} \left\{ \hat{F}_\tau(l) < \hat{F}_\tau(k) + \frac{1}{n} \right\} + \sum_{l=1, l \geq k}^K \hat{p}_\tau(l) \mathbb{1} \left\{ \hat{F}_\tau(l) < \hat{F}_\tau(k) \right\}. \end{aligned}$$

If  $1 \leq l < k$  then  $\hat{F}_\tau(l) \leq \hat{F}_\tau(k) < \hat{F}_\tau(k) + \frac{1}{n}$ , and if  $K \geq l \geq k$  then  $\hat{F}_\tau(l) \geq \hat{F}_\tau(k)$ . The above equation can be rewritten as

$$\hat{G}_\tau(k) = \sum_{l=1, l < k}^K \hat{p}_\tau(l) = \hat{F}_\tau(k-1).$$

Going back to trying to find an expression for  $\hat{C}_{\alpha;\tau}$

$$k \in \hat{C}_{\alpha;\tau} \iff \hat{F}_\tau(k-1) < (1-\alpha) \left( 1 + \frac{1}{n} \right)$$

$$\begin{aligned} &\Longleftrightarrow k - 1 \leq \hat{Q}_\tau \left( (1 - \alpha) \left( 1 + \frac{1}{n} \right) \right) \\ &\Longleftrightarrow k \leq \hat{Q}_\tau \left( (1 - \alpha) \left( 1 + \frac{1}{n} \right) \right) + 1, \end{aligned}$$

where the empirical marginal upper quantile function is defined as, for every level  $a \in [0, 1]$ ,

$$\hat{Q}_\tau^{\text{up}}(a) := \max \left\{ k \in \{1, \dots, K\} : \hat{F}_\tau < a. \right\} \quad (6)$$

the empirical marginal cumulative distribution function  $\hat{F}_\tau$  in Equation (Eq. (5)).  $\square$

**Proposition 3** (Instant-by-instant prediction lower band, MDist-full). *Let  $\xi_1, \dots, \xi_n$  be  $n$  independent copies of  $\xi$ . For any control level  $\alpha \in [0, 1]$ , the band  $\hat{B}_\alpha^{\text{MDist-full,lo}}$  defined as, for every  $\tau \in \{1, \dots, T\}$*

$$\hat{B}_{\alpha;\tau}^{\text{MDist-full,lo}} := \left\{ k \in \{1, \dots, K\} : k \geq \hat{Q}_\tau^{\text{lo}} \left( \alpha \left( 1 + \frac{1}{n} \right) - \frac{1}{n} \right) \right\} \quad (7)$$

*ensures control of the instant-by-instant coverage of  $\xi$  at a control level  $\alpha$ , where the marginal empirical lower quantile function  $\hat{Q}_\tau^{\text{lo}}$  is defined in Equation (Eq. (8)).*

*Proof of Proposition 3.* Let  $\alpha \in [0, 1]$  and  $\tau \in \{1, \dots, T\}$ . A full conformal prediction region  $\hat{C}_{\alpha;\tau}$  can be defined as

$$\hat{C}_{\alpha;\tau} := \{k \in \{1, \dots, K\} : \hat{\pi}(k) > \alpha\},$$

where

$$\hat{\pi}(k) := \frac{1 + \sum_{i=1}^n \mathbb{1} \left\{ \hat{F}_\tau^k(\xi_{i;\tau}) \leq \hat{F}_\tau^k(k) \right\}}{n + 1}.$$

Since  $\xi_{1;\tau}, \dots, \xi_{n;\tau}$  are i.i.d. copies of  $\xi_\tau$ ,  $\hat{C}_{\alpha;\tau}$  is a confidence prediction region (Vovk, Gamerman, and Shafer, 2005)

$$\mathbb{P} \left[ \xi_\tau \in \hat{C}_{\alpha;\tau} \right] \geq 1 - \alpha.$$

Let us now provide an explicit expression of  $\hat{C}_{\alpha;\tau}$ .

$$\begin{aligned} k \in \hat{C}_{\alpha;\tau} &\Longleftrightarrow \hat{\pi}_\tau(k) > \alpha \\ &\Longleftrightarrow \frac{1 + \sum_{i=1}^n \mathbb{1} \left\{ \hat{F}_\tau^k(\xi_{i;\tau}) \leq \hat{F}_\tau^k(k) \right\}}{n + 1} > \alpha \\ &\Longleftrightarrow \frac{1}{n} \sum_{i=1}^n \mathbb{1} \left\{ \hat{F}_\tau^k(\xi_{i;\tau}) \leq \hat{F}_\tau^k(k) \right\} > \alpha \left( 1 + \frac{1}{n} \right) - \frac{1}{n}. \end{aligned}$$

Let us focus on the l.h.s. term. Let us define the function  $\tilde{G}_\tau : \{1, \dots, K\} \rightarrow [0, 1]$  such that for every  $k$

$$\begin{aligned} \tilde{G}_\tau(k) &:= \frac{1}{n} \sum_{i=1}^n \mathbb{1} \left\{ \hat{F}_\tau^k(\xi_{i;\tau}) \leq \hat{F}_\tau^k(k) \right\} \\ &= \frac{1}{n} \sum_{i=1}^n \sum_{l=1}^K \mathbb{1} \{l = \xi_{i;\tau}\} \mathbb{1} \left\{ \hat{F}_\tau^k(\xi_{i;\tau}) \leq \hat{F}_\tau^k(k) \right\} \end{aligned}$$

$$\begin{aligned}
&= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1} \left\{ \hat{F}_\tau^k(l) \leq \hat{F}_\tau^k(k) \right\} \\
&= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1} \left\{ \frac{n}{n+1} \hat{F}_\tau(l) + \frac{1}{n+1} \mathbb{1} \{k \leq l\} \leq \frac{n}{n+1} \hat{F}_\tau(k) + \frac{1}{n+1} \right\} \\
&= \sum_{l=1}^K \hat{p}_\tau(l) \mathbb{1} \left\{ \hat{F}_\tau(l) + \frac{1}{n} \mathbb{1} \{k \leq l\} \leq \hat{F}_\tau(k) + \frac{1}{n} \right\} \\
&= \sum_{l=1, l \leq k}^K \hat{p}_\tau(l) \mathbb{1} \left\{ \hat{F}_\tau(l) \leq \hat{F}_\tau(k) + \frac{1}{n} \right\} + \sum_{l=1, l > k}^K \hat{p}_\tau(l) \mathbb{1} \left\{ \hat{F}_\tau(l) \leq \hat{F}_\tau(k) \right\}.
\end{aligned}$$

If  $1 \leq l < k$  then  $\hat{F}_\tau(l) \leq \hat{F}_\tau(k) < \hat{F}_\tau(k) + \frac{1}{n}$ , and if  $K \geq l > k$  and  $\hat{p}_\tau(l) \neq 0$ , then

$$\hat{F}_\tau(l) = \sum_{\tilde{l}=1}^l \hat{p}_\tau(\tilde{l}) = \sum_{\tilde{l}=1}^k \hat{p}_\tau(\tilde{l}) + \sum_{\tilde{l}=k+1}^l \hat{p}_\tau(\tilde{l}) = \hat{F}_\tau(k) + \sum_{\tilde{l}=k+1}^l \hat{p}_\tau(\tilde{l}) > \hat{F}_\tau(k),$$

since  $\sum_{\tilde{l}=k+1}^l \hat{p}_\tau(\tilde{l})$  includes  $\hat{p}_\tau(l) \neq 0$ . Hence, the function  $\tilde{G}_\tau$  can be rewritten as

$$\tilde{G}_\tau(k) = \sum_{l=1, l \leq k}^K \hat{p}_\tau(l) = \hat{F}_\tau(k).$$

Going back to trying to find an expression for  $\hat{C}_{\alpha;\tau}$

$$\begin{aligned}
k \in \hat{C}_{\alpha;\tau} &\iff \hat{F}_\tau(k) > \alpha \left( 1 + \frac{1}{n} \right) - \frac{1}{n} \\
&\iff k \geq \tilde{Q}_\tau \left( \alpha \left( 1 + \frac{1}{n} \right) - \frac{1}{n} \right),
\end{aligned}$$

where the empirical marginal lower quantile function is defined as, for every level  $a \in (0, 1)$

$$\tilde{Q}_\tau^{\text{lo}}(a) := \min \left\{ k \in \{1, \dots, K\} : \hat{F}_\tau(k) > a \right\}, \quad (8)$$

the empirical marginal cumulative distribution function  $\hat{F}_\tau$  in Equation (Eq. (5)).  $\square$

**Proposition 4** (Instant-by-instant prediction band, MDist-full). *Let  $\xi_1, \dots, \xi_n$  be  $n$  independent copies of  $\xi$ . For any control level  $\alpha \in [0, 1]$ , and for any  $\alpha_{\text{lo}}, \alpha_{\text{up}} \in [0, 1]$  such that  $\alpha_{\text{lo}} + \alpha_{\text{up}} = \alpha$ , the band  $\hat{B}_\alpha^{\text{MDist-full}}$  defined as, for every  $\tau \in \{1, \dots, T\}$*

$$\begin{aligned}
\hat{B}_{\alpha;\tau}^{\text{MDist-full}} &:= \left\{ k \in \{1, \dots, K\} : \hat{Q}_\tau^{\text{lo}} \left( \alpha_{\text{lo}} \left( 1 + \frac{1}{n} \right) - \frac{1}{n} \right) \leq k, \right. \\
&\quad \left. k \leq \hat{Q}_\tau^{\text{up}} \left( (1 - \alpha_{\text{up}}) \left( 1 + \frac{1}{n} \right) \right) + 1 \right\}, \quad (9)
\end{aligned}$$

ensures control of the instant-by-instant coverage of  $\xi$  at a control level  $\alpha$ .

*Proof of Proposition 4.* Let  $\alpha \in [0, 1]$  and  $\tau \in \{1, \dots, T\}$ . It follows from union bound that

$$\begin{aligned}
\mathbb{P} \left[ \xi_\tau \in \hat{B}_{\alpha;\tau}^{\text{MDist-full}} \right] &= \mathbb{P} \left[ \xi_\tau \in \hat{B}_{\alpha_{\text{up}};\tau}^{\text{MDist-full,up}} \cap \hat{B}_{\alpha_{\text{lo}};\tau}^{\text{MDist-full,lo}} \right] \\
&= 1 - \mathbb{P} \left[ \xi_\tau \in \left( \hat{B}_{\alpha_{\text{up}};\tau}^{\text{MDist-full,up}} \right)^c \cup \left( \hat{B}_{\alpha_{\text{lo}};\tau}^{\text{MDist-full,lo}} \right)^c \right]
\end{aligned}$$

$$\begin{aligned}
&\geq 1 - \left( \mathbb{P} \left[ \xi_\tau \in \left( \hat{B}_{\alpha_{\text{up}};\tau}^{\text{MDist-full,up}} \right)^c \right] + \mathbb{P} \left[ \xi_\tau \in \left( \hat{B}_{\alpha_{\text{lo}};\tau}^{\text{MDist-full,lo}} \right)^c \right] \right) \\
&\geq 1 - (\alpha_{\text{up}} + \alpha_{\text{lo}}) = 1 - \alpha.
\end{aligned}$$

□

**Proposition 5** (Instant-by-instant prediction band, MDist-Split). *Let  $\xi_1, \dots, \xi_n$  be  $n$  independent copies of  $\xi$ . Let  $I_1$  and  $I_2$  be two index sets with cardinal  $n_1$  and  $n_2$  respectively such that  $I_1 \sqcup I_2 = \{1, \dots, n\}$ . For any control level  $\alpha \in [0, 1]$ , the band  $\hat{B}_\alpha^{\text{MDist-split}}$  defined as, for every  $\tau \in \{1, \dots, T\}$*

$$\hat{B}_{\alpha;\tau}^{\text{MDist-split}} := \left\{ k \in \{1, \dots, K\} : \tilde{Q}_{D_1;\tau}^{\text{lo}} \left( \hat{t}_{\alpha;\tau}^{\text{MDist-split}} \right) \leq k \leq \tilde{Q}_{D_1;\tau}^{\text{up}} \left( 1 - \hat{t}_{\alpha;\tau}^{\text{MDist-split}} \right) \right\}, \quad (10)$$

*ensures control of the instant-by-instant coverage of  $\xi$  at a control level  $\alpha$ , where the threshold  $\hat{t}_{\alpha;\tau}^{\text{MDist-split}}$  is defined in Equation (Eq. (12)) and the marginal empirical upper and lower quantile function  $\tilde{Q}_{D_1;\tau}^{\text{up}}$  and  $\tilde{Q}_{D_1;\tau}^{\text{lo}}$  are defined in Equations (Eq. (13)) and (Eq. (14)) respectively.*

*Proof of Proposition 5.* Let  $\tau \in \{1, \dots, T\}$ . Define the empirical marginal cumulative density function  $\hat{F}_{D_1;\tau}$  trained on the data set  $D_1 := \{\xi_i, i \in I_1\}$  as, for every  $k \in \{1, \dots, K\}$

$$\hat{F}_{D_1;\tau}(k) := \frac{1}{n_1} \sum_{i \in I_1} \mathbb{1} \{ \xi_{i;\tau} \leq k \}. \quad (11)$$

Upon the dataset  $D_2 := \{\xi_i, i \in I_2\}$ , one can define split conformal prediction region  $\hat{C}_{\alpha;\tau}$  as

$$\hat{C}_{\alpha;\tau} := \{ k \in \{1, \dots, K\} : \hat{\pi}_{D_2;\tau}(k) > \alpha \},$$

where for every  $k \in \{1, \dots, K\}$

$$\hat{\pi}_{D_2;\tau}(k) := \frac{1 + \sum_{i \in I_2} \mathbb{1} \{ \hat{A}_{D_1;\tau}(\xi_{i;\tau}) \leq \hat{A}_{D_1;\tau}(k) \}}{1 + n_2},$$

where the conformity score  $\hat{A}_{D_1;\tau}(k) := \min \left( \hat{F}_{D_1;\tau}(k), 1 - \hat{F}_{D_1;\tau}(k) \right)$ . Since  $\xi_{1;\tau}, \dots, \xi_{n_2;\tau}$  are i.i.d. copies of  $\xi_\tau$ , it follows that  $\hat{C}_{\alpha;\tau}$  is a confidence prediction region

$$\mathbb{P} \left( \xi_\tau \in \hat{C}_{\alpha;\tau} \right) \geq 1 - \alpha.$$

Let us now compute an explicit expression for  $\hat{C}_{\alpha;\tau}$ .

$$\begin{aligned}
&k \in \hat{C}_{\alpha;\tau} \\
&\iff \frac{1 + \sum_{i \in I_2} \mathbb{1} \{ \hat{A}_{D_1;\tau}(\xi_{i;\tau}) \leq \hat{A}_{D_1;\tau}(k) \}}{1 + n_2} > \alpha \\
&\iff \frac{1}{n_2} \sum_{i \in I_2} \mathbb{1} \{ \hat{A}_{D_1;\tau}(\xi_{i;\tau}) \leq \hat{A}_{D_1;\tau}(k) \} > \alpha \left( 1 + \frac{1}{n_2} \right) - \frac{1}{n_2} \\
&\iff \hat{A}_{D_1;\tau}(k) \geq \hat{\ell}_{\alpha;\tau},
\end{aligned}$$

where the level  $\hat{\ell}_{\alpha;\tau}^{\text{MDist-split}}$  is defined as

$$\hat{\ell}_{\alpha;\tau}^{\text{MDist-split}} := \inf \left\{ t \in [0, 1] : \frac{1}{n_2} \sum_{i \in I_2} \mathbb{1} \{ \hat{A}_{D_1;\tau}(\xi_{i;\tau}) \leq t \} > \alpha \left( 1 + \frac{1}{n_2} \right) - \frac{1}{n_2} \right\}$$

$$= \hat{A}_{D_1;\tau} \left( \xi_{(i_{n_2;\alpha}),\tau} \right), \quad (12)$$

where the index  $i_{n_2;\alpha} := \lceil \alpha(n_2 + 1) - 1 \rceil$ , and  $\hat{A}_{D_1;\tau}(\xi_{(1);\tau}) \leq \dots \leq \hat{A}_{D_1;\tau}(\xi_{(n_2);\tau})$  are the conformity scores sorted in increasing order. Going back to trying to find an expression of the prediction region

$$\begin{aligned} k &\in \hat{C}_{\alpha;\tau} \\ \iff \min \left( \hat{F}_{D_1;\tau}(k), 1 - \hat{F}_{D_1;\tau}(k) \right) &\geq \hat{t}_{\alpha;\tau}^{\text{MDist-split}}, \\ \iff \hat{F}_{D_1;\tau}(k) &\geq \hat{t}_{\alpha;\tau}^{\text{MDist-split}}, \text{ and } 1 - \hat{F}_{D_1;\tau}(k) \geq \hat{t}_{\alpha;\tau}^{\text{MDist-split}}, \\ \iff \hat{t}_{\alpha;\tau}^{\text{MDist-split}} &\leq \hat{F}_{D_1;\tau}(k) \leq 1 - \hat{t}_{\alpha;\tau}^{\text{MDist-split}}, \\ \iff \tilde{Q}_{D_1;\tau}^{\text{lo}} \left( \hat{t}_{\alpha;\tau}^{\text{MDist-split}} \right) &\leq k \leq \tilde{Q}_{D_1;\tau}^{\text{up}} \left( 1 - \hat{t}_{\alpha;\tau}^{\text{MDist-split}} \right), \end{aligned}$$

where the empirical marginal upper quantile function built from  $D_1$  is defined as, for every level  $a \in [0, 1]$  is defined as

$$\tilde{Q}_{D_1;\tau}^{\text{up}}(1 - a) := \max \left\{ k \in \{1, \dots, K\} : \hat{F}_{D_1;\tau}(k) \leq 1 - a \right\}, \quad (13)$$

as for the lower one

$$\tilde{Q}_{D_1;\tau}^{\text{lo}}(a) := \min \left\{ k \in \{1, \dots, K\} : \hat{F}_{D_1;\tau}(k) \geq a \right\}. \quad (14)$$

□

**Proposition 6** ( $\gamma$ -simultaneous band). *Let  $\xi_1, \dots, \xi_n$  be i.i.d. copies of  $\xi$ . Let  $I_1$  and  $I_2$  be two index sets with cardinal  $n_1$  and  $n_2$  respectively such that  $I_1 \sqcup I_2 = \{1, \dots, n\}$ . Let  $\hat{A}_{D_1;\tau} : \{1, \dots, K\} \rightarrow \mathbb{R}$ ,  $k \mapsto \hat{A}_{D_1;\tau}(k)$  be a conformity-measure over  $\{1, \dots, K\}$  built from the dataset  $D_1 := \{\xi_i : i \in I_1\}$ .*

*For any slack level  $\gamma \in [0, 1]$ , control level  $\alpha \in [0, 1]$ , the prediction band  $\hat{B}_{\gamma;\alpha}$  defined as for every  $\tau \in \{1, \dots, T\}$*

$$\hat{B}_{\gamma;\alpha;\tau} := \left\{ k \in \{1, \dots, K\} : \hat{A}_{D_1;\tau}(k) \geq \hat{t}_{D_1;\gamma} \left( \xi_{(i_{n_2;\alpha})} \right) \right\}$$

*ensures control of the  $\gamma$ -simultaneous coverage of  $\xi$  at a level  $\alpha$  where the cut-off level  $\hat{t}_{D_1;\gamma} \left( \xi_{(i_{n_2;\alpha})} \right)$  is defined in Equation (Eq. 16) from the conformity score defined in Equation (Eq. 15).*

*Proof of Proposition 6.* Let  $\gamma \in [0, 1]$  and  $\alpha \in [0, 1]$ . One can define a conformity measure  $\hat{t}_{D_1;\gamma} : \{1, \dots, K\}^T \rightarrow \mathbb{R}$  such that for every  $\xi \in \{1, \dots, K\}^T$

$$\hat{t}_{D_1;\gamma}(\xi) := \sup \left\{ t \in \mathbb{R} : \frac{1}{T} \sum_{\tau=1}^T \mathbb{1} \left\{ \hat{A}_{D_1;\tau}(\xi_\tau) \geq t \right\} \geq 1 - \gamma \right\} = \hat{A}_{D_1;(\tau_{\gamma,T})} \left( \xi_{(\tau_{\gamma,T})} \right), \quad (15)$$

where  $\hat{A}_{D_1;(1)}(\xi_{(1)}) \geq \dots \geq \hat{A}_{D_1;(T)}(\xi_{(T)})$  are sorted in decreasing order and  $\tau_{\gamma,T} := \lceil T(1 - \gamma) \rceil$ . Upon the dataset  $D_2 := \{\xi_i : i \in I_2\}$ , one can define a split conformal prediction region as

$$\hat{C}_{\gamma;\alpha} := \left\{ \xi \in \{1, \dots, K\}^T : \hat{\pi}_{D_2}(\xi) > \alpha \right\},$$

where for every  $\xi \in \{1, \dots, K\}^T$

$$\hat{\pi}_{D_2}(\xi) := \frac{1 + \sum_{i \in I_2} \left\{ \hat{t}_{D_1;\gamma}(\xi_i) \leq \hat{t}_{D_1;\gamma}(\xi) \right\}}{n_2 + 1}.$$



Let us now provide an explicit expression for  $\hat{C}_{\gamma;\alpha}$ ,

$$\begin{aligned}\xi \in \hat{C}_{\gamma;\alpha} &\iff \frac{1 + \sum_{i \in I_2} \{\hat{t}_{D_1;\gamma}(\xi_i) \leq \hat{t}_{D_1;\gamma}(\xi)\}}{n_2 + 1} > \alpha \\ &\iff \frac{1}{n_2} \sum_{i \in I_2} \{\hat{t}_{D_1;\gamma}(\xi_i) \leq \hat{t}_{D_1;\gamma}(\xi)\} > \left(1 + \frac{1}{n_2}\right) \alpha - \frac{1}{n_2} \\ &\iff \hat{t}_{D_1;\gamma}(\xi) \geq \hat{t}_{D_1;\gamma}(\xi_{(i_{n_2;\alpha})}),\end{aligned}\tag{16}$$

where  $\hat{t}_{D_1;\gamma}(\xi_{(i_{n_2;\alpha})}) \leq \dots \leq \hat{t}_{D_1;\gamma}(\xi_{(i_{n_2;\alpha})})$  are sorted in increasing order and  $i_{n_2;\alpha} := \lceil (n_2 + 1)\alpha - 1 \rceil$ . Hence, the prediction band  $\hat{B}_{\gamma;\alpha}$  defined as for every  $\tau \in \{1, \dots, T\}$

$$\hat{B}_{\gamma;\alpha;\tau} := \left\{k \in \{1, \dots, K\} : \hat{A}_{D_1;\tau}(k) \geq \hat{t}_{D_1;\gamma}(\xi_{(i_{n_2;\alpha})})\right\}$$

ensures the following

$$\begin{aligned}\mathbb{P} \left[ \frac{1}{T} \sum_{\tau=1}^T \mathbb{1} \left\{ \xi_\tau \in \hat{B}_{\gamma;\alpha;\tau} \right\} \right] &= \mathbb{P} \left[ \frac{1}{T} \sum_{\tau=1}^T \mathbb{1} \left\{ \hat{A}_{D_1;\tau}(\xi_\tau) \geq \hat{t}_{D_1;\gamma}(\xi_{(i_{n_2;\alpha})}) \right\} \right] \\ &= \mathbb{P} \left[ \hat{t}_{D_1;\gamma}(\xi) \geq \hat{t}_{D_1;\gamma}(\xi_{(i_{n_2;\alpha})}) \right] \\ &= \mathbb{P} \left[ \xi \in \hat{C}_{\gamma;\alpha} \right] \geq 1 - \alpha.\end{aligned}$$

□

**Corollary 1** ( $\gamma$ -simultaneous band, MD-Split). *Let  $\xi_1, \dots, \xi_n$  be i.i.d. copies of  $\xi$ . Let  $I_1$  and  $I_2$  be two index sets with cardinal  $n_1$  and  $n_2$  respectively such that  $I_1 \sqcup I_2 = \{1, \dots, n\}$ .*

*For any slack  $\gamma \in (0, 1)$ , control level  $\alpha \in [0, 1]$ , the prediction band  $\hat{B}_{\gamma;\alpha}^{\text{MD-split}}$  such that for every  $\tau \in \{1, \dots, T\}$*

$$\hat{B}_{\gamma;\alpha;\tau}^{\text{MD-split}} := \left\{k \in \{1, \dots, K\} : \hat{p}_{D_1;\tau}(k) \geq \hat{\ell}_{\gamma;\alpha}^{\text{MD-Split}}\right\},\tag{17}$$

*ensures control of  $\gamma$ -simultaneous coverage of  $\xi$  at a control level  $\alpha$ , where empirical marginal probability density function  $\hat{p}_{D_1;\tau}$  is defined in Equation (Eq. (18)), and the threshold is  $\hat{\ell}_{\gamma;\alpha}^{\text{MD-Split}}$  is defined in Equation (Eq. (19)).*

*Proof of Corollary 1.* For every  $\tau \in \{1, \dots, K\}$ , define empirical marginal probability density function  $\hat{p}_{D_1;\tau}$  built from the dataset  $D_1 := \{\xi_i : i \in I_1\}$  as, for every  $k \in \{1, \dots, K\}$

$$\hat{p}_{D_1;\tau}(k) := \frac{1}{n_1} \sum_{i \in I_1} \mathbb{1} \{\xi_{i;\tau} = k\}.\tag{18}$$

Applying Proposition 6 by choosing for every  $\tau \in \{1, \dots, T\}$ , and every  $k \in \{1, \dots, K\}$ ,  $\hat{A}_{D_1;\tau}(k) := \hat{p}_{D_1;\tau}(k)$ , the prediction band  $\hat{B}_{\gamma;\alpha}^{\text{MD-split}}$  defined as, for every  $\tau \in \{1, \dots, T\}$

$$\begin{aligned}\hat{B}_{\gamma;\alpha;\tau}^{\text{MD-split}} &:= \left\{k \in \{1, \dots, K\} : \hat{A}_{D_1;\tau}(k) \geq \hat{t}_{D_1;\gamma}(\xi_{(i_{n_2;\alpha})})\right\} \\ &= \left\{k \in \{1, \dots, K\} : \hat{p}_{D_1;\tau}(k) \geq \hat{\ell}_{\gamma;\alpha}^{\text{MD-Split}}\right\},\end{aligned}$$

ensures control of  $\gamma$ -simultaneous coverage of  $\xi$  at a control level  $\alpha$ , where the threshold  $\hat{\ell}_{\gamma;\alpha}^{\text{MD-Split}}$  is defined as

$$\hat{\ell}_{\gamma;\alpha}^{\text{MD-Split}} := \hat{t}_{D_1;\gamma}(\xi_{(i_{n_2;\alpha})}).\tag{19}$$

□

**Corollary 2** ( $\gamma$ -simultaneous band, MHPD-Split). *Let  $\xi_1, \dots, \xi_n$  be i.i.d. copies of  $\xi$ . Let  $I_1$  and  $I_2$  be two index sets with cardinal  $n_1$  and  $n_2$  respectively such that  $I_1 \sqcup I_2 = \{1, \dots, n\}$ .*

*For any slack  $\gamma \in (0, 1)$ , control level  $\alpha \in [0, 1]$ , the prediction band  $\hat{B}_{\gamma; \alpha}^{\text{MHPD-split}}$  such that for every  $\tau \in \{1, \dots, T\}$*

$$\hat{B}_{\gamma; \alpha; \tau}^{\text{MHPD-split}} := \left\{ k \in \{1, \dots, K\} : \hat{p}_{D_1; \tau}(k) > \hat{\ell}_{\gamma; \alpha; \tau}^{\text{MHPD-split}} \right\}, \quad (20)$$

*ensures  $\gamma$ -simultaneous coverage of  $\xi$  at a control level  $\alpha$ , where empirical marginal probability density function  $\hat{p}_{D_1; \tau}$  is defined in Equation (Eq. (18)), and the threshold  $\hat{\ell}_{\gamma; \alpha; \tau}^{\text{MHPD-split}}$  is defined in Equation (Eq. (22)).*

*Proof of Corollary 2.* For every  $\tau \in \{1, \dots, K\}$ , and every  $k \in \{1, \dots, K\}$  define the rank  $\hat{R}_{D_1; \tau}(k)$  of  $k$  at the instant  $\tau$  built from the dataset  $D_1 := \{\xi_i : i \in I_1\}$  as

$$\hat{R}_{D_1; \tau}(k) := \sum_{l=1}^K \hat{p}_{D_1; \tau}(l) \mathbb{1} \{ \hat{p}_{D_1; \tau}(l) \leq \hat{p}_{D_1; \tau}(k) \}. \quad (21)$$

Applying Proposition 6 by choosing for every  $\tau \in \{1, \dots, T\}$ , and every  $k \in \{1, \dots, K\}$ ,  $\hat{A}_{D_1; \tau}(k) := \hat{R}_{D_1; \tau}(k)$ , the prediction band  $\hat{B}_{\gamma; \alpha}^{\text{MD-split}}$  defined as, for every  $\tau \in \{1, \dots, T\}$

$$\begin{aligned} \hat{B}_{\gamma; \alpha; \tau}^{\text{MHPD-split}} &:= \left\{ k \in \{1, \dots, K\} : \hat{A}_{D_1; \tau}(k) \geq \hat{t}_{D_1; \gamma} \left( \xi_{(i_{n_2; \alpha})} \right) \right\} \\ &= \left\{ k \in \{1, \dots, K\} : \hat{R}_{D_1; \tau}(k) \geq \hat{t}_{D_1; \gamma} \left( \xi_{(i_{n_2; \alpha})} \right) \right\} \\ &= \left\{ k \in \{1, \dots, K\} : \hat{p}_{D_1; \tau}(k) > \hat{\ell}_{\gamma; \alpha; \tau}^{\text{MHPD-split}} \right\}, \end{aligned}$$

ensures control of  $\gamma$ -simultaneous coverage of  $\xi$  at a control level  $\alpha$ , where the cut-off  $\hat{\ell}_{\gamma; \alpha; \tau}^{\text{MHPD-split}}$  is defined as

$$\hat{\ell}_{\gamma; \alpha; \tau}^{\text{MHPD-split}} := \sup \left\{ \ell \in [0, 1] : \sum_{l=1}^K \hat{p}_{D_1; \tau}(l) \mathbb{1} \{ \hat{p}_{D_1; \tau}(l) > \ell \} > 1 - \hat{t}_{D_1; \gamma}^{\text{MHPD-split}} \left( \xi_{(i_{n_2; \alpha})} \right) \right\}, \quad (22)$$

□

**Corollary 3** ( $\gamma$ -simultaneous band, MDist-Split). *Let  $\xi_1, \dots, \xi_n$  be  $n$  independent copies of  $\xi$ . Let  $I_1$  and  $I_2$  be two index sets with cardinal  $n_1$  and  $n_2$  respectively such that  $I_1 \sqcup I_2 = \{1, \dots, n\}$ .*

*For any slack  $\gamma \in (0, 1)$ , any control level  $\alpha \in [0, 1]$ , the band  $\hat{B}_{\gamma; \alpha}^{\text{MDist-split}}$  defined as, for every  $\tau \in \{1, \dots, T\}$*

$$\hat{B}_{\gamma; \alpha; \tau}^{\text{MDist-split}} := \left\{ k \in \{1, \dots, K\} : \hat{Q}_{D_1, \tau}^{\text{lo}} \left( \hat{t}_{\gamma; \alpha}^{\text{MDist-split}} \right) \leq k \leq \hat{Q}_{D_1, \tau}^{\text{lo}} \left( 1 - \hat{t}_{\gamma; \alpha}^{\text{MDist-split}} \right) \right\}, \quad (23)$$

*ensures  $\gamma$ -simultaneous coverage of  $\xi$  at a control level  $\alpha$ , where the level  $\hat{t}_{\gamma; \alpha}^{\text{MDist-split}}$  is defined in Equation (Eq. (25)).*

*Proof of Corollary 3.* Applying Proposition 6, for every  $\tau \in \{1, \dots, K\}$ , and  $k \in \{1, \dots, K\}$

$$\hat{A}_{D_1; \tau}(k) := \min \left( \hat{F}_{D_1; \tau}(k), 1 - \hat{F}_{D_1; \tau}(k) \right), \quad (24)$$

the prediction band  $\hat{B}_{\gamma; \alpha}^{\text{MDist-split}}$  defined as, for every  $\tau \in \{1, \dots, T\}$

$$\hat{B}_{\gamma; \alpha; \tau}^{\text{MHPD-split}} := \left\{ k \in \{1, \dots, K\} : \hat{A}_{D_1; \tau}(k) \geq \hat{t}_{D_1; \gamma} \left( \xi_{(i_{n_2; \alpha})} \right) \right\}$$

$$\begin{aligned}
&= \left\{ k \in \{1, \dots, K\} : \min \left( \hat{F}_{D_1; \tau}(k), 1 - \hat{F}_{D_1; \tau}(k) \right) \geq \hat{t}_{\gamma; \alpha}^{\text{MDist-split}} \right\} \\
&= \left\{ k \in \{1, \dots, K\} : \hat{Q}_{D_1, \tau}^{\text{lo}} \left( \hat{t}_{\gamma; \alpha}^{\text{MDist-split}} \right) \leq k \leq \hat{Q}_{D_1, \tau}^{\text{lo}} \left( 1 - \hat{t}_{\gamma; \alpha}^{\text{MDist-split}} \right) \right\},
\end{aligned}$$

ensures control of  $\gamma$ -simultaneous coverage of  $\xi$  at a control level  $\alpha$ , where the level  $\hat{t}_{\gamma; \alpha}^{\text{MDist-split}}$  is defined as

$$\hat{t}_{\gamma; \alpha}^{\text{MDist-split}} := \hat{t}_{D_1; \gamma} \left( \xi_{(i_{n_2; \alpha})} \right). \quad (25)$$

□

## References

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