

ROTATION SAMPLING FOR FUNCTIONAL DATA

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Supplementary Material

S1 Proof of Proposition 1

It suffices to prove the proposition at each replacement time τ_r , $r = 0, \dots, m$. To do so, we proceed by induction on r . For $r = 0$, the property is true by definition of SRSWOR. Assume that the property holds at rank $(r - 1)$ for some $0 < r < m$. Fix the stratum U_h and consider a subset $D \subset U_h$ of size $n_h(\tau_r)$. In order to establish the property at rank r , we must show that

$$P(s_h(\tau_r) = D) = \left(\frac{N_h}{n_h(\tau_r)} \right)^{-1}. \quad (\text{S1.1})$$

By the total probability formula and the induction assumption,

$$\begin{aligned} P(s_h(\tau_r) = D) &= \sum_{\substack{D' \subset U_h \\ \#(D')=n_h(\tau_{r-1})}} P(s_h(\tau_r) = D | s_h(\tau_{r-1}) = D') P(s_h(\tau_{r-1}) = D') \\ &= \left(\frac{N_h}{n_h(\tau_{r-1})} \right)^{-1} \sum_{\substack{D' \subset U_h \\ \#(D')=n_h(\tau_{r-1})}} P(s_h(\tau_r) = D | s_h(\tau_{r-1}) = D'). \end{aligned} \quad (\text{S1.2})$$

We seek the subsets $D' \subset U_h$ of size $n_h(\tau_{r-1})$ such that $P(s_h(\tau_r) = D | s_h(\tau_{r-1}) = D') > 0$. Let D' be such a subset and let $k = \#(D \cap D')$. For the sample $s_h(\tau_{r-1}) = D'$ to transform into $s_h(\tau_r) = D$, the $(n_h(\tau_{r-1}) - k)$ units in $D' \setminus D$ must be removed from $s_h(\tau_{r-1})$ and the $(n_h(\tau_r) - k)$ units in $D \setminus D'$ must be added to $s_h(\tau_{r-1})$. This entails that $k = (1 - \alpha_h)n_h(\tau_{r-1})$. Reciprocally, any subset $D' \subset U_h$ of size $n_h(\tau_{r-1})$ verifying the condition $\#(D \cap D') = (1 - \alpha_h)n_h(\tau_{r-1})$ can be transformed in D with the above operations. This condition is thus necessary and sufficient and the number $d_h(\tau_r)$ of subsets $D' \subset U_h$ of size $n_h(\tau_{r-1})$ satisfying it is

$$d_h(\tau_r) = \binom{n_h(\tau_r)}{(1 - \alpha_h)n_h(\tau_{r-1})} \binom{N_h - n_h(\tau_r)}{n_h(\tau_{r-1}) - (1 - \alpha_h)n_h(\tau_{r-1})}, \quad (\text{S1.3})$$

where the first factor accounts for the possible choices of the $(1 - \alpha_h)n_h(\tau_{r-1})$ common elements between D and D' and the second factor accounts for the possible choices of the $(n_h(\tau_r) - (1 - \alpha_h)n_h(\tau_{r-1}))$ remaining elements of D' in $U_h \setminus D$.

For each of the previous subsets, the properties of SRSWOR imply that

$$P(s_h(\tau_r) = D | s_h(\tau_{r-1}) = D') = \left[\binom{N_h - n_h(\tau_{r-1})}{n_h(\tau_r) - (1 - \alpha_h)n_h(\tau_{r-1})} \binom{n_h(\tau_{r-1})}{\alpha_h n_h(\tau_{r-1})} \right]^{-1}. \quad (\text{S1.4})$$

Plugging (S1.3)-(S1.4) in (S1.2), one deduces (S1.1), which completes the induction. \square

S2 Proof of Lemma 1

Let $D_0, \dots, D_m \subset D$. Fix $r \in \{1, \dots, m\}$. We will establish that

$$\begin{aligned} P(s_h(\tau_r) \cap D = D_r | s_h(\tau_{r-1}) \cap D = D_{r-1}, \dots, s_h(\tau_0) \cap D = D_0) \\ = P(s_h(\tau_r) \cap D = D_r | s_h(\tau_{r-1}) \cap D = D_{r-1}). \end{aligned} \quad (\text{S2.1})$$

For $1 \leq i \leq r$, we express $\{s_h(\tau_i) \cap D = D_i\}$ more conveniently as $\{s_h(\tau_i) \in A_i\}$, where $A_i = \{a_i \in \mathcal{P}(U_h) : \#a_i = n_h(\tau_i), a_i \supset D_i\}$. ($\mathcal{P}(U_h)$ is the set of all subsets of U_h). We denote by \mathbf{a} the generic element $(a_{r-1}, \dots, a_0) \in A_{r-1} \times \dots \times A_0$. Let $A = \{\mathbf{a} \in A_{r-1} \times \dots \times A_0 : P(\mathbf{a}) > 0\}$. The Markov property of $\{s_h(\tau_0), \dots, s_h(\tau_m)\}$ yields

$$\begin{aligned} P(s_h(\tau_r) \in A_r | s_h(\tau_{r-1}) \in A_{r-1}, \dots, s_h(\tau_0) \in A_0) \\ = \frac{\sum_{\mathbf{a} \in A} P(s_h(\tau_r) \in A_r | s_h(\tau_{r-1}) = a_{r-1}) P(s_h(\tau_{r-1}) = a_{r-1}, \dots, s_h(\tau_0) = a_0)}{\sum_{\mathbf{a} \in A} P(s_h(\tau_{r-1}) = a_{r-1}, \dots, s_h(\tau_0) = a_0)}. \end{aligned} \quad (\text{S2.2})$$

Invoking again the Markov property, we obtain that

$$\begin{aligned} P(s_h(\tau_{r-1}) = a_{r-1}, \dots, s_h(\tau_0) = a_0) \\ = P(s_h(\tau_{r-1}) = a_{r-1} | s_h(\tau_{r-2}) = a_{r-2}) \times \dots \\ \times P(s_h(\tau_1) = a_1 | s_h(\tau_0) = a_0) \times P(s_h(\tau_0) = a_0) \end{aligned} \quad (\text{S2.3})$$

for all $\mathbf{a} \in A$.

Equation (S1.4) and the properties of SRSWOR show that (S2.3) only depends on $\alpha_h, n_h(\tau_0), \dots, n_h(\tau_r)$ and N_h . As a consequence, (S2.2) rewrites as

$$\begin{aligned} P(s_h(\tau_r) \in A_r | s_h(\tau_{r-1}) \in A_{r-1}, \dots, s_h(\tau_0) \in A_0) \\ = \frac{1}{\#A} \sum_{\mathbf{a} \in A} P(s_h(\tau_r) \in A_r | s_h(\tau_{r-1}) = a_{r-1}). \end{aligned} \quad (\text{S2.4})$$

Each collection A_i describes the value of the sample $s_h(\tau_i)$ in D , namely D_i , and the unspecified possible values of $s_h(\tau_i)$ outside of D . In view of Proposition 1, it is easily seen that the conditional probability distribution of $s_h(\tau_{r-1}) \setminus D$ given that $s_h(\tau_{r-1}) \cap D = D_{r-1}$ coincides with the SRSWOR of $n_h(\tau_{r-1}) - d_{r-1}$ units in $U_h \setminus D$. A combinatorial consequence is that

$$\#\{(a_{r-2}, \dots, a_0) \in A_{r-2} \times \dots \times A_0 : (a_{r-1}, \dots, a_0) \in A\} = \frac{\#A}{\#A_{r-1}} \quad (\text{S2.5})$$

for all $a_{r-1} \in A_{r-1}$. By following the proof of Proposition 1, one could explicitly find $(\#A/\#A_{r-1})$ in terms of α_h, N_h and the $n_h(\tau_i), d_i, \#(D_i \cap D_{i+1}), 0 \leq i \leq r-1$.

Combining (S2.4) and (S2.5), we finally obtain

$$\begin{aligned} P(s_h(\tau_r) \in A_r | s_h(\tau_{r-1}) \in A_{r-1}, \dots, s_h(\tau_0) \in A_0) \\ &= \frac{1}{\#A_{r-1}} \sum_{a_{r-1} \in A_{r-1}} P(s_h(\tau_r) \in A_r | s_h(\tau_{r-1}) = a_{r-1}) \\ &= P(s_h(\tau_r) \in A_r | s_h(\tau_{r-1}) \in A_{r-1}). \quad \square \end{aligned}$$

S3 Proof of Proposition 2

We decompose the studied sum as $\sum_{\ell=1}^4 A_\ell(t, t')$, where

$$A_\ell(t, t') = \sum_{\substack{i, j, k, l \in U_h \\ \mathcal{C}_{ijkl} = \ell}} \mathbb{E}(I_i(t)I_j(t)I_k(t')I_l(t')) \tilde{X}_i(t)\tilde{X}_j(t)\tilde{X}_k(t')\tilde{X}_l(t')$$

and $\mathcal{C}_{ijkl} = \#\{i, j, k, l\}$. Hereafter, we compute $\mathbb{E}(I_i(t)I_j(t)I_k(t')I_l(t'))$ using the properties of SRSWOR and we compute sums involving $\tilde{X}_i(t)\tilde{X}_j(t)\tilde{X}_k(t')\tilde{X}_l(t')$ using the identity $\sum_{k \in U_h} \tilde{X}_k(t) = 0$. Let i^*, j^*, j^*, l^* be four distinct units in U_h .

We begin with the straightforward calculation of $A_1(t, t')$:

$$A_1(t, t') = \mathbb{E}(I_{i^*}(t)I_{i^*}(t')) \sum_k \tilde{X}_k^2(t)\tilde{X}_k^2(t'). \quad (\text{S3.1})$$

The term $A_2(t, t')$ can be expressed as

$$\begin{aligned} A_2(t, t') &= \mathbb{E}(I_{i^*}(t)I_{k^*}(t')) \sum_{i \neq k} \tilde{X}_i^2(t)\tilde{X}_k^2(t') \\ &\quad + 2 \mathbb{E}(I_{i^*}(t)I_{i^*}(t')I_{k^*}(t)I_{k^*}(t')) \sum_{i \neq l} \tilde{X}_i(t)\tilde{X}_i(t')\tilde{X}_l(t)\tilde{X}_l(t') \\ &\quad + 2 \mathbb{E}(I_{i^*}(t)I_{i^*}(t')I_{k^*}(t')) \sum_{i \neq k} \tilde{X}_i^2(t)\tilde{X}_i(t')\tilde{X}_k(t') \\ &\quad + 2 \mathbb{E}(I_{i^*}(t)I_{k^*}(t)I_{k^*}(t')) \sum_{i \neq k} \tilde{X}_i(t)\tilde{X}_k(t)\tilde{X}_k^2(t'), \end{aligned}$$

that is,

$$\begin{aligned}
A_2(t, t') &= \mathbb{E}(I_{i^*}(t)I_{k^*}(t')) \left[(N_h - 1)^2 \gamma_h(t, t) \gamma_h(t', t') - \sum_{k \in U_h} \tilde{X}_k^2(t) \tilde{X}_k^2(t') \right] \\
&+ 2 \mathbb{E}(I_{i^*}(t)I_{i^*}(t')I_{k^*}(t)I_{k^*}(t')) \left[(N_h - 1)^2 \gamma_h^2(t, t') - \sum_{k \in U_h} \tilde{X}_k^2(t) \tilde{X}_k^2(t') \right] \\
&- 2 \left[\mathbb{E}(I_{i^*}(t)I_{i^*}(t')I_{k^*}(t')) + \mathbb{E}(I_{i^*}(t)I_{k^*}(t)I_{k^*}(t')) \right] \sum_{k \in U_h} \tilde{X}_k^2(t) \tilde{X}_k^2(t').
\end{aligned} \tag{S3.2}$$

Next, we have

$$\begin{aligned}
A_3(t, t') &= \mathbb{E}(I_{i^*}(t)I_{j^*}(t)I_{k^*}(t')) \sum_{i \neq j \neq k} \tilde{X}_i(t) \tilde{X}_j(t) \tilde{X}_k^2(t') \\
&+ \mathbb{E}(I_{i^*}(t)I_{k^*}(t')I_{l^*}(t')) \sum_{i \neq k \neq l} \tilde{X}_i^2(t) \tilde{X}_k(t') \tilde{X}_l(t') \\
&+ 4 \mathbb{E}(I_{i^*}(t)I_{i^*}(t')I_{j^*}(t)I_{k^*}(t')) \sum_{i \neq j \neq k} \tilde{X}_{i^*}(t) \tilde{X}_{i^*}(t) \tilde{X}_j(t) \tilde{X}_k(t')
\end{aligned}$$

and a further expansion yields

$$\begin{aligned}
A_3(t, t') &= \left[\mathbb{E}(I_{i^*}(t)I_{j^*}(t)I_{k^*}(t')) + \mathbb{E}(I_{i^*}(t)I_{k^*}(t')I_{l^*}(t')) \right] \\
&\times \left[-(N_h - 1)^2 \gamma_h(t, t) \gamma_h(t', t') + 2 \sum_{k \in U_h} \tilde{X}_k^2(t) \tilde{X}_k^2(t') \right] \\
&+ 4 \mathbb{E}(I_{i^*}(t)I_{i^*}(t')I_{j^*}(t)I_{k^*}(t')) \\
&\times \left[-(N_h - 1)^2 \gamma_h^2(t, t') + 2 \sum_{k \in U_h} \tilde{X}_k^2(t) \tilde{X}_k^2(t') \right].
\end{aligned} \tag{S3.3}$$

To compute $A_4(t, t')$, recall that $\sum_{i, j, k, l \in U_h} \tilde{X}_i(t) \tilde{X}_j(t) \tilde{X}_k(t') \tilde{X}_l(t') = 0$ and use the decomposition

$$\sum_{i, j, k, l \in U_h} \tilde{X}_i(t) \tilde{X}_j(t) \tilde{X}_k(t') \tilde{X}_l(t') = \sum_{\ell=1}^4 \sum_{\substack{i, j, k, l \in U_h \\ \mathcal{C}_{ijkl} = \ell}} \tilde{X}_i(t) \tilde{X}_j(t) \tilde{X}_k(t') \tilde{X}_l(t')$$

together with the expressions of $A_1(t, t')$, $A_2(t, t')$, $A_3(t, t')$ to obtain

$$\begin{aligned}
A_4(t, t') &= \mathbb{E}(I_{i^*}(t)I_{j^*}(t)I_{k^*}(t')I_{l^*}(t')) \times \\
&\left[(N_h - 1)^2 \gamma_h(t, t) \gamma_h(t', t') + 2(N_h - 1)^2 \gamma_h^2(t, t') - 6 \sum_{k \in U_h} \tilde{X}_k^2(t) \tilde{X}_k^2(t') \right].
\end{aligned} \tag{S3.4}$$

The proof is completed by gathering (S3.1)–(S3.4) and observing that all terms involving $\sum_{k \in U_h} \tilde{X}_k^2(t) \tilde{X}_k^2(t')$ are of lower order N_h thanks to (A1). \square