## Residual resampling with multinomial residuals

Suzie Brown

29 April 2021

THIS DOCUMENT IS OBSOLETE. For a collection of failed attempts, including this one, see resmn\_roundup\_210526. —SB

(A1) The conditional distribution of parental indices  $a_t^{(1:N)}$  given offspring counts  $\nu_t^{(1:N)}$  is uniform over all assignments such that  $|\{j:a_t^{(j)}=i\}|=\nu_t^{(i)}$  for all i.

Corollary 1. Consider an SMC algorithm using residual resampling with multinomial residuals, such that (A1) is satisfied. Assume that there exists a constant  $a \in [1, \infty)$  such that for all x, x', t,

$$\frac{1}{a} \le g_t(x, x') \le a.$$

Assume that  $\mathbb{P}[\tau_N(t) = \infty] = 0$  for all finite t. Let  $(G_t^{(n,N)})_{t\geq 0}$  denote the genealogy of a random sample of n terminal particles from the output of the algorithm when the total number of particles used is N. Then, for any fixed n, the time-scaled genealogy  $(G_{\tau_N(t)}^{(n,N)})_{t\geq 0}$  converges to Kingman's n-coalescent as  $N \to \infty$ , in the sense of finite-dimensional distributions.

*Proof.* With residual-multinomial resampling, for each i

$$\nu_t^{(i)} \mid w_t^{(1:N)} \stackrel{d}{=} |Nw_t^{(i)}| + X_i$$

where  $X_i \sim \text{Binomial}(R, r_i)$ . As usual,  $R := N - \sum_{i=1}^N \lfloor Nw_t^{(i)} \rfloor$  and  $r_i := (Nw_t^{(i)} - \lfloor Nw_t^{(i)} \rfloor)/R$ . If R = 0 then  $r_i = 0$  for all i and the following calculations remain correct.—SB We can therefore compute

$$\begin{split} \mathbb{E}[(\nu_t^{(i)})_2 \mid w_t^{(1:N)}] &= \mathbb{E}\left[(\lfloor Nw_t^{(i)} \rfloor + X_i)(\lfloor Nw_t^{(i)} \rfloor + X_i - 1) \mid w_t^{(1:N)}\right] \\ &= (\lfloor Nw_t^{(i)} \rfloor)_2 + 2\lfloor Nw_t^{(i)} \rfloor E[X_i \mid w_t^{(1:N)}] + \mathbb{E}[(X_i)_2 \mid w_t^{(1:N)}] \\ &= (\lfloor Nw_t^{(i)} \rfloor)_2 + 2\lfloor Nw_t^{(i)} \rfloor Rr_i + (R)_2 r_i^2 \end{split}$$

Suzie Brown 1

using the moments of the Binomial distribution. We also have

$$\begin{split} \mathbb{E}[(\nu_t^{(i)})_3 \mid w_t^{(1:N)}] &= \mathbb{E}\left[ (\lfloor Nw_t^{(i)} \rfloor + X_i) (\lfloor Nw_t^{(i)} \rfloor + X_i - 1) (\lfloor Nw_t^{(i)} \rfloor + X_i - 2) \, \Big| \, w_t^{(1:N)} \right] \\ &= \lfloor Nw_t^{(i)} \rfloor^3 + \lfloor Nw_t^{(i)} \rfloor^2 \mathbb{E}[3X_i - 3 \mid w_t^{(1:N)}] \\ &+ \lfloor Nw_t^{(i)} \rfloor \mathbb{E}[X_i(X_i - 1) + X_i(X_i - 2) + (X_i - 1)(X_i - 2) \mid w_t^{(1:N)}] \\ &+ \mathbb{E}[(X_i)_3 \mid w_t^{(1:N)}] \\ &= \lfloor Nw_t^{(i)} \rfloor^3 - 3 \lfloor Nw_t^{(i)} \rfloor^2 + 3 \lfloor Nw_t^{(i)} \rfloor^2 \mathbb{E}[X_i \mid w_t^{(1:N)}] \\ &+ \lfloor Nw_t^{(i)} \rfloor \mathbb{E}[3X_i^2 - 6X_i + 2 \mid w_t^{(1:N)}] + E[(X_i)_3 \mid w_t^{(1:N)}] \\ &= \left( \lfloor Nw_t^{(i)} \rfloor^3 - 3 \lfloor Nw_t^{(i)} \rfloor^2 + 2 \lfloor Nw_t^{(i)} \rfloor \right) + 3 \left( \lfloor Nw_t^{(i)} \rfloor^2 - \lfloor Nw_t^{(i)} \rfloor \right) \mathbb{E}[X_i \mid w_t^{(1:N)}] \\ &+ 3 \lfloor Nw_t^{(i)} \rfloor \mathbb{E}[(X_i)_2 \mid w_t^{(1:N)}] + \mathbb{E}[(X_i)_3 \mid w_t^{(1:N)}] \\ &= (\lfloor Nw_t^{(i)} \rfloor)_3 + 3 (\lfloor Nw_t^{(i)} \rfloor)_2 Rr_i + 3 \lfloor Nw_t^{(i)} \rfloor (R)_2 r_i^2 + (R)_3 r_i^3 \\ &\leq \left( \lfloor Nw_t^{(i)} \rfloor + Rr_i \right) \left\{ (\lfloor Nw_t^{(i)} \rfloor)_2 + 2 \lfloor Nw_t^{(i)} \rfloor Rr_i + (R)_2 r_i^2 \right\} \\ &= Nw_t^{(i)} \mathbb{E}[(\nu_t^{(i)})_2 \mid w_t^{(1:N)}] \\ &\leq a^2 \mathbb{E}[(\nu_t^{(i)})_2 \mid w_t^{(1:N)}], \end{split}$$

using the almost sure bound  $w_t^{(i)} \leq a^2/N$ .

...

To complete the proof we need to exchange the conditioning on  $w_t^{(1:N)}$  for conditioning on  $\mathcal{H}_t$  so we can then invoke the D-separation and tower property to get

$$\frac{1}{(N)_3} \sum_{i=1}^{N} \mathbb{E}_t[(\nu_t^{(i)})_3] \le b_N \frac{1}{(N)_2} \sum_{i=1}^{N} \mathbb{E}_t[(\nu_t^{(i)})_2]$$

for some sequence  $b_N \to 0$ . Perhaps we should expect bounds on  $q_t$  to be required, so that  $\varepsilon$  will also appear in  $b_N$ . —SB

Suzie Brown 2