Fix  $\varepsilon > 0$ . Let N be large enough that  $\varepsilon > 1/N$ , and let  $A_i(r) := \{\nu_r^{(i)} \le N\varepsilon\}$ . Following [Möhle and Sagitov, 2003, Proof of Lemma 5.5],

$$\begin{split} &\frac{1}{N(N)_2} \sum_{i=1}^N (\nu_r^{(i)})_2 \Bigg[ \nu_r^{(i)} + \frac{1}{N} \sum_{j \neq i} (\nu_r^{(j)})^2 \Bigg] \mathbbm{1}_{A_i(r)} \\ & \leq \frac{1}{N(N)_2} \sum_{i=1}^N (\nu_r^{(i)})_2 \Bigg[ N \varepsilon + \frac{1}{N} \sum_{j=1}^N (\nu_r^{(j)})_2 + \frac{1}{N} \sum_{j=1}^N \nu_r^{(j)} \Bigg] \mathbbm{1}_{A_i(r)} \\ & \leq \Bigg[ \varepsilon c_N(r) + \frac{1}{N} c_N(r) + \frac{N \varepsilon}{N^2(N)_2} \sum_{i=1}^N \nu_r^{(i)} \sum_{j=1}^N (\nu_r^{(j)})_2 \Bigg] \mathbbm{1}_{A_i(r)} = \Bigg( 2\varepsilon + \frac{1}{N} \Bigg) c_N(r), \end{split}$$

and

$$\frac{1}{N(N)_2} \sum_{i=1}^N (\nu_r^{(i)})_2 \Bigg[ \nu_r^{(i)} + \frac{1}{N} \sum_{j \neq i} (\nu_r^{(j)})^2 \Bigg] \mathbbm{1}_{A_i(r)^c} \leq \sum_{i=1}^N \mathbbm{1}_{A_i(r)^c}.$$

Thus

$$\mathbb{E}\left[\sum_{r=\tau_{N}(t_{0})+1}^{\tau_{N}(t_{k})}\mathbb{1}_{\{c_{N}(r)\leq\binom{n-2}{2}D_{N}(r)\}}\right]\leq\mathbb{E}\left[\sum_{r=\tau_{N}(t_{0})+1}^{\tau_{N}(t_{k})}\mathbb{1}_{\left\{\binom{n-2}{2}^{-1}-2\varepsilon-1/N\right)\leq\sum_{i=1}^{N}\mathbb{1}_{A_{i}(r)^{c}/c_{N}(r)}\right\}\right]$$

where the ratio  $\mathbb{1}_{A_i(r)^c}/c_N(r)$  is well defined because

$$A_i(r)^c \Rightarrow c_N(r) = \frac{1}{(N)_2} \sum_{i=1}^N (\nu_t^{(i)})_2 \ge \frac{\varepsilon(N\varepsilon - 1)}{N - 1} = \varepsilon \left(\varepsilon - \frac{1}{N}\right) > 0.$$

Hence also

$$\mathbb{E}\left[\sum_{r=\tau_{N}(t_{0})+1}^{\tau_{N}(t_{k})}\mathbb{1}_{\{c_{N}(r)\leq\binom{n-2}{2}D_{N}(r)\}}\right]\leq\mathbb{E}\left[\sum_{r=\tau_{N}(t_{0})+1}^{\tau_{N}(t_{k})}\mathbb{1}_{\left\{\left(\binom{n-2}{2}\right)^{-1}-2\varepsilon-1/N\right)\varepsilon(\varepsilon-1/N)\leq\sum_{i=1}^{N}\mathbb{1}_{A_{i}(r)^{c}}\right\}\right],$$

whereupon [Koskela et al., 2018, Lemma 2] and the conditional Markov inequality yield

$$\mathbb{E}\left[\sum_{r=\tau_{N}(t_{0})+1}^{\tau_{N}(t_{k})} \mathbb{1}_{\{c_{N}(r) \leq {n-2 \choose 2}D_{N}(r)\}}\right]$$

$$\leq \frac{1}{\left({n-2 \choose 2}^{-1} - 2\varepsilon - 1/N\right)\varepsilon(\varepsilon - 1/N)} \mathbb{E}\left[\sum_{r=\tau_{N}(t_{0})+1}^{\tau_{N}(t_{k})} \sum_{i=1}^{N} \mathbb{1}_{A_{i}(r)^{c}}\right].$$

A further two invocations of [Koskela et al., 2018, Lemma 2] with the conditional Markov inequality in between result in

$$\mathbb{E}\left[\sum_{r=\tau_{N}(t_{0})+1}^{\tau_{N}(t_{k})} \mathbb{1}_{\{c_{N}(r) \leq \binom{n-2}{2}D_{N}(r)\}}\right] \\
\leq \frac{1}{\left(\binom{n-2}{2}^{-1} - 2\varepsilon - 1/N\right)\varepsilon(\varepsilon - 1/N)} \mathbb{E}\left[\sum_{r=\tau_{N}(t_{0})+1}^{\tau_{N}(t_{k})} \sum_{i=1}^{N} \frac{(\nu_{r}^{(i)})_{3}}{(N\varepsilon)_{3}}\right] \\
\leq \frac{N(N)_{2}}{\left(\binom{n-2}{2}^{-1} - 2\varepsilon - 1/N\right)\varepsilon(\varepsilon - 1/N)(N\varepsilon)_{3}} \mathbb{E}\left[\sum_{r=\tau_{N}(t_{0})+1}^{\tau_{N}(t_{k})} D_{N}(r)\right] \\
\to \frac{1}{\left(\binom{n-2}{2}^{-1} - 2\varepsilon\right)\varepsilon^{5}} \times 0$$

by [Koskela et al., 2018, Eq. (3)] as  $N \to \infty$ , as required.

## References

- Jere Koskela, Paul A Jenkins, Adam M Johansen, and Dario Spanò. Asymptotic genealogies of interacting particle systems with an application to sequential Monte Carlo. arXiv:1804.01811, 2018.
- M. Möhle and S. Sagitov. Coalescent patterns in exchangeable diploid population models.  $Journal\ of\ Mathematical\ Biology,\ 47:337-352,\ 2003.$