Non-triviality condition (fuller details of paper appendix)

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The following theorem will be used in each section. It is a filtered version of the second Borel–Cantelli lemma, which can be found for instance in Durrett (2019, Theorem 4.3.4).

Lemma 1. Let $(\mathcal{F}_t)_{t\geq 0}$ be a filtration with $\mathcal{F}_0 = \{\emptyset, \Omega\}$. Let $(B_t)_{t\geq 0}$ be a sequence of events such that $B_t \in \mathcal{F}_t$ for all t. Then the events $\{B_t \ i.o.\}$ and $\{\sum_{t=1}^{\infty} \mathbb{P}[B_t \mid \mathcal{F}_{t-1}] = \infty\}$ are almost surely equal.

We will also use the following equivalence in each section.

Lemma 2. Let τ_N denote the generalised inverse of c_N , i.e.

$$\tau_N(t) = \min \left\{ s \ge 1 : \sum_{r=1}^s c_N(r) \ge t \right\}.$$

Suppose that there exists $N_0 \in \mathbb{N}$ such that almost surely for all $N > N_0$, $c_N(t)$ is bounded away from zero for infinitely many t. Then for all $N > N_0$, for all finite t, $\mathbb{P}[\tau_N(t) = \infty] = 0$.

Proof. Applying the definition of $\tau_N(t)$,

$$\mathbb{P}[\tau_N(t) = \infty] = 0 \Leftrightarrow \mathbb{P}[\tau_N(t) < \infty] = 1$$

$$\Leftrightarrow \mathbb{P}\left[\min\left\{s \ge 1 : \sum_{r=1}^s c_N(r) \ge t\right\} < \infty\right] = 1$$

$$\Leftrightarrow \mathbb{P}\left[\exists s < \infty : \sum_{r=1}^s c_N(r) \ge t\right] = 1$$

A sufficient condition for the last line is that, almost surely for all $N > N_0$, $c_N(r)$ is bounded away from zero for infinitely many r.

Combining Lemmata 1 and 2, we see that a sufficient condition for non-triviality is

$$\sum_{t=0}^{\infty} \mathbb{P}[c_N(t) > 2/N^2 \mid \mathcal{F}_{t-1}] = \infty.$$
 (1)

Multinomial resampling

Proposition 1. In standard SMC with multinomial resampling, there exists N_0 such that for all $N > N_0$, for all finite t, $\mathbb{P}[\tau_N(t) = \infty] = 0$.

Proof. We have the following lower bound:

$$\mathbb{E}[c_N(t) \mid \mathcal{F}_{t-1}] \ge \frac{\varepsilon^4}{Na^4}.$$

Since $c_N(t) \in [0,1]$ almost surely, for any fixed N the "worst-case" distribution of $c_N(t)$ (i.e. maximising $\mathbb{P}[c_N(t) = 0 \mid \mathcal{F}_{t-1}]$) is two atoms, at 0 and 1. To ensure the correct expectation, the atom at 1 must have weight $\mathbb{E}[c_N(t) \mid \mathcal{F}_{t-1}]$, which is bounded below by the above inequality. Hence for any finite N,

$$\sum_{t=0}^{\infty} \mathbb{P}[c_N(t) > 2/N^2 \mid \mathcal{F}_{t-1}] \ge \sum_{t=0}^{\infty} \frac{\varepsilon^4}{Na^4} = \infty.$$

Applying Lemmata 1 and 2 yields the result.

Conditional SMC with multinomial resampling

Proposition 2. In conditional SMC with multinomial resampling, there exists N_0 such that for all $N > N_0$, for all finite t, $\mathbb{P}[\tau_N(t) = \infty] = 0$.

Proof. We have the following lower bound:

$$\mathbb{E}[c_N(t) \mid \mathcal{F}_{t-1}] \ge \frac{1}{(N)_2} \left\{ \frac{2\varepsilon^2}{a^2} + \frac{(N)_3 \varepsilon^4}{(N-1)^2 a^4} \right\}.$$

Since $c_N(t) \in [0, 1]$ almost surely, for any fixed N the "worst-case" distribution of $c_N(t)$ (i.e. maximising $\mathbb{P}[c_N(t) = 0 \mid \mathcal{F}_{t-1}]$) is two atoms, at 0 and 1. To ensure the correct expectation, the atom at 1 must have weight $\mathbb{E}[c_N(t) \mid \mathcal{F}_{t-1}]$, which is bounded below by the above inequality. Hence for any finite N,

$$\sum_{t=0}^{\infty} \mathbb{P}[c_N(t) > 2/N^2 \mid \mathcal{F}_{t-1}] \ge \sum_{t=0}^{\infty} \frac{1}{(N)_2} \left\{ \frac{2\varepsilon^2}{a^2} + \frac{(N)_3 \varepsilon^4}{(N-1)^2 a^4} \right\} = \infty.$$

Applying Lemmata 1 and 2 yields the result.

Stochastic rounding

The stochastic rounding case is more involved than the others, because there is not a positive lower bound on $\mathbb{E}_t[c_N(t)]$ that holds for all weight vectors. (In particular, $\mathbb{E}_t[c_N(t)] = 0$ when the weights are all exactly equal.) First notice the following.

Lemma 3. Let $\mathbf{w} = (w_1, \dots, w_N) \in \mathcal{S}_{N-1}$ and resample by stochastic rounding.

- (i) If $w_i \geq 2/N$ for some i, then $\mathbb{P}[c_N(t) > 2/N^2 \mid \mathcal{H}_t] = 1$.
- (ii) If $w_i = 0$ for some i, then $\mathbb{P}[c_N(t) > 2/N^2 \mid \mathcal{H}_t] = 1$.

Proof. For any N it is easy to see, by considering the possible values of $c_N(t)$, that $c_N(t) > 2/N^2$ if and only if $c_N(t) \neq 0$. The only way to attain $c_N(t) = 0$ is to assign exactly one offspring to each particle. In case (i) particle i is assigned at least two offspring, so $c_N(t)$ cannot be equal to zero. In case (ii) particle i is assigned zero offspring, so $c_N(t)$ cannot be equal to zero.

The upshot of Lemma 3 is that we need only prove (1) in the case where all weights are in (0, 2/N), since it holds automatically otherwise.

Lemma 4. Define $\mathbf{w}^{\delta} := \frac{1}{N}\{(1,\ldots,1) + \delta\mathbf{e}_i - \delta\mathbf{e}_j\}$ for any $i \neq j$ and $0 < \delta < 1$. Then $\mathbb{P}[c_N(t) > 2/N^2 | \mathcal{H}_t, \mathbf{w}_t = \mathbf{w}_{\delta}] > \delta\varepsilon^3$.

Proof. We use a bound on $\mathbb{P}[\nu_t^{(i)} = \lfloor Nw_t^{(i)} \rfloor]$ from the proof of Corollary 1 in the draft paper:

$$\mathbb{P}[\nu_t^{(i)} = \lfloor Nw_t^{(i)} \rfloor \mid \mathcal{H}_t] =: p_0 = 1 - p_1 \le 1 - (Nw_t^{(i)} - \lfloor Nw_t^{(i)} \rfloor) \varepsilon^{(2\lfloor Nw_t^{(i)} \rfloor + 1)}.$$

Then

$$\mathbb{P}[c_N(t) \leq 2/N^2 | \mathcal{H}_t, \mathbf{w}_t = \mathbf{w}_{\delta}] = \mathbb{P}[\nu_t^{(1:N)} = (1, \dots, 1) | \mathcal{H}_t, \mathbf{w}_t = \mathbf{w}_{\delta}]$$

$$= \mathbb{P}[\nu_t^{(i)} = 1, \nu_t^{(j)} = 1 | \mathcal{H}_t, \mathbf{w}_t = \mathbf{w}_{\delta}]$$

$$= \mathbb{P}[\nu_t^{(i)} = 1 | \mathcal{H}_t, \mathbf{w}_t = \mathbf{w}_{\delta}]$$

$$\leq 1 - (Nw_{\delta}^{(i)} - \lfloor Nw_{\delta}^{(i)} \rfloor) \varepsilon^{(2\lfloor Nw_{\delta}^{(i)} \rfloor + 1)}$$

$$= 1 - \{N(1 + \delta)/N - 1\} \varepsilon^3$$

$$= 1 - \delta \varepsilon^3.$$

since the offspring counts are deterministically equal to one apart from particles i and j, and it remains that $\nu_t^{(i)} = 1$ if and only if $\nu_t^{(j)} = 1$.

Lemma 5. For any $\delta \in (0,1)$, denote $S_{N-1}^{\delta} := \{ \mathbf{w} \in S_{N-1} : \forall i, 0 < w_i < \frac{2}{N}; \max_i w_i \ge \frac{1+\delta}{N} \}$. Then for all $\mathbf{w} \in S_{N-1}^{\delta}$, $\mathbb{P}[c_N(t) > 2/N^2 | \mathbf{w}] \ge \mathbb{P}[c_N(t) > 2/N^2 | \mathbf{w}_{\delta}]$.

Proof. Fix arbitrary $\mathbf{w} \in \mathcal{S}_{N-1}^{\delta}$. Let i^* be then index of the particle with the largest weight. Denote $\mathcal{I} := \{i \in \{1, \dots, N\} : w_i > 1/N\}$. Notice that

$$\mathbb{P}[c_N(t) \le 2/N^2 | \mathbf{w}] = \mathbb{P}[\nu_t^{(i)} = 1 \,\forall i \in \{1, \dots, N\} | \mathbf{w}] = \mathbb{P}[\nu_t^{(i)} = 1 \,\forall i \in \mathcal{I} | \mathbf{w}].$$

This is true because all weights are in (0,2/N), so for $i \in \mathcal{I}, \nu_t^{(i)} \in \{1,2\}$, and for $i \notin \mathcal{I}, \nu_t^{(i)} \in \{0,1\}$; and the offspring counts must sum to N (a generalisation of the argument used in Lemma 4).

We can then decompose this probability into a product of conditional probabilities:

$$\begin{split} \mathbb{P}[\nu_t^{(i)} &= 1 \, \forall i \in \mathcal{I} | \mathbf{w}] = \prod_{i \in \mathcal{I}} \mathbb{P}[\nu_t^{(i)} = 1 | \nu_t^{(j)} = 1 \, \forall j < i \in \mathcal{I}; \mathbf{w}] \\ &= \mathbb{P}[\nu_t^{(i^*)} = 1 | \mathbf{w}] \prod_{i \neq i^* \in \mathcal{I}} \mathbb{P}[\nu_t^{(i)} = 1 | \nu_t^{(i^*)} = 1; \nu_t^{(j)} = 1 \, \forall j < i \in \mathcal{I}; \mathbf{w}] \\ &\leq \mathbb{P}[\nu_t^{(i^*)} = 1 | \mathbf{w}]. \end{split}$$

The last line is equal to the probability $\mathbb{P}[c_N(t) \leq 2/N^2|\mathbf{w}]$ in the case where $|\mathcal{I}| = 1$, i.e. the only weight larger than 1/N is w_{i^*} .

In other words, $\mathbb{P}[c_N(t) > 2/N^2 | \mathbf{w}]$ is minimised on $\mathcal{S}_{N-1}^{\delta}$ by having only one weight larger than 1/N, in which case the values of the other weights do not affect this probability.

We therefore find that a minimum of $\mathbb{P}[c_N(t) > 2/N^2 | \mathbf{w}]$ on $\mathcal{S}_{N-1}^{\delta}$ is given by $\mathbf{w}_{\delta'}$, for some $\delta' \geq \delta$. It only remains to show that taking $\delta' > \delta$ does not decrease the probability. This is a consequence of Lemma 4, where we see that $\mathbb{P}[c_N(t) > 2/N^2 | \mathbf{w}_{\delta'}]$ is monotonically increasing in δ' . Thus the minimum of $\mathbb{P}[c_N(t) > 2/N^2 | \mathbf{w}]$ is attained at $\mathbf{w} = \mathbf{w}_{\delta}$, as required. (Although this minimum is not unique, we have shown explicitly that it is a global minimum on $\mathcal{S}_{N-1}^{\delta}$.)

Theorem 1. Consider a sequential Monte Carlo algorithm using any stochastic rounding as its resampling scheme. If there exists $\mu > 0$ such that $\mathbb{P}\{\max_i w_t^{(i)} \geq (1+\delta)/N \mid \mathcal{H}_t\} \geq \mu$ for infinitely many t then $\mathbb{P}\{\tau_N(t) = \infty\} = 0$ for all N > 1 and for all finite t.

Proof. Combining Lemmata [7–9] we see that, for any $\mathbf{w} \in \mathcal{S}_{N-1}$ such that $\max_i w_i \geq \frac{1+\delta}{N}$, we have the bound $\mathbb{P}[c_N(t) > 2/N^2 | \mathbf{w}] \geq \delta \varepsilon^3$. By the law of total probability,

$$\mathbb{P}[c_N(t) > 2/N^2 \mid \mathcal{H}_t] \ge \mathbb{P}[c_N(t) > 2/N^2 \mid \mathcal{H}_t, \max w_i \ge (1+\delta)/N] \, \mathbb{P}[\max w_i \ge (1+\delta)/N \mid \mathcal{H}_t] \ge \mu \delta \varepsilon^3$$

for those infinitely many t where $\mathbb{P}\{\max_i w_t^{(i)} \geq (1+\delta)/N \mid \mathcal{H}_t\} \geq \mu$. Using the D-separation established in [draft paper, Cor 1 proof], we can write

$$\mathbb{P}[c_{N}(t) > 2/N^{2} \mid \mathcal{F}_{t-1}] = \mathbb{E}[\mathbb{I}_{\{c_{N}(t) > 2/N^{2}\}} \mid \mathcal{F}_{t-1}]
= \mathbb{E}[\mathbb{E}[\mathbb{I}_{\{c_{N}(t) > 2/N^{2}\}} \mid \mathcal{H}_{t}] \mid \mathcal{F}_{t-1}]
= \mathbb{E}[\mathbb{P}[c_{N}(t) > 2/N^{2} \mid \mathcal{H}_{t}] \mid \mathcal{F}_{t-1}].$$

Hence this probability is bounded below by $\mu\delta\varepsilon^3$ for infinitely many t. We therefore have

$$\sum_{t=0}^{\infty} \mathbb{P}[c_N(t) > 2/N^2 \mid \mathcal{F}_{t-1}] \ge \sum_{j=0}^{\infty} \mu \delta \varepsilon^3 = \infty, \tag{2}$$

and applying Theorem [1 - that BC2 statement], almost surely $c_N(t) > 2/N^2$ for infinitely many t. As argued in Lemma [2], this is sufficient for the result.

The lemma below is here to clear up any uncertainty about the tower property / D-separation argument, as used in this proof in the paper.

Lemma 6. Let A, B be events such that A is measurable with respect to \mathcal{F}_t , and B is measurable with respect to \mathcal{H}_t (but not vice versa), and neither event is measurable with respect to \mathcal{F}_{t-1} . (In the real proof we have $A := \{c_N(t) > 2/N^2\}$ and $B := \{\max_i w_t^{(i)} - \min_i w_t^{(i)} \ge 2\delta/N\}$). Then

$$\mathbb{P}[A \mid \mathcal{F}_{t-1}, B] = \mathbb{E}[\mathbb{P}[A \mid \mathcal{H}_t] \mid \mathcal{F}_{t-1}, B]$$
(3)

and

$$\mathbb{P}[A \mid \mathcal{H}_t] \ge \mathbb{P}[A \mid \mathcal{H}_t, B] \mathbb{I}_B. \tag{4}$$

Proof. For the first point,

$$\begin{split} \mathbb{P}[A \mid \mathcal{F}_{t-1}, B] &= \mathbb{E}[\mathbb{I}_A \mid \mathcal{F}_{t-1}, B] = \frac{\mathbb{E}[\mathbb{I}_A \mathbb{I}_B \mid \mathcal{F}_{t-1}]}{\mathbb{P}[B \mid \mathcal{F}_{t-1}]} = \frac{\mathbb{E}[\mathbb{E}[\mathbb{I}_A \mathbb{I}_B \mid \mathcal{H}_t, \mathcal{F}_{t-1}] \mid \mathcal{F}_{t-1}]}{\mathbb{P}[B \mid \mathcal{F}_{t-1}]} = \frac{\mathbb{E}[\mathbb{E}[\mathbb{I}_A \mathbb{I}_B \mid \mathcal{H}_t] \mid \mathcal{F}_{t-1}]}{\mathbb{P}[B \mid \mathcal{F}_{t-1}]} \\ &= \frac{\mathbb{E}[\mathbb{E}[\mathbb{I}_A \mid \mathcal{H}_t] \mathbb{I}_B \mid \mathcal{F}_{t-1}]}{\mathbb{P}[B \mid \mathcal{F}_{t-1}]} = \frac{\mathbb{E}[\mathbb{E}[\mathbb{I}_A \mid \mathcal{H}_t] \mid \mathcal{F}_{t-1}, B] \mathbb{P}[B \mid \mathcal{F}_{t-1}]}{\mathbb{P}[B \mid \mathcal{F}_{t-1}]} \\ &= \mathbb{E}[\mathbb{E}[\mathbb{I}_A \mid \mathcal{H}_t] \mid \mathcal{F}_{t-1}, B] = \mathbb{E}[\mathbb{P}[A \mid \mathcal{H}_t] \mid \mathcal{F}_{t-1}, B]. \end{split}$$

For the second point,

$$\mathbb{P}[A \mid \mathcal{H}_t] = \mathbb{P}[A \mid \mathcal{H}_t, B] \mathbb{P}[B \mid \mathcal{H}_t] + \mathbb{P}[A \mid \mathcal{H}_t, B^c] \mathbb{P}[B^c \mid \mathcal{H}_t] \ge \mathbb{P}[A \mid \mathcal{H}_t, B] \mathbb{P}[B \mid \mathcal{H}_t] = \mathbb{P}[A \mid \mathcal{H}_t, B] \mathbb{I}_B$$

since B is \mathcal{H}_t -measurable.

The next Lemma shows how these two results are helpful in our scenario of Corollary 1.

Lemma 7. Let $A := \{c_N(t) > 2/N^2\}$. Let $B := \{\max_i w_t^{(i)} - \min_i w_t^{(i)} \ge 2\delta/N\}$. (Notice that these events satisfy the measurability properties in the previous Lemma.) As an assumption in Corollary 1 we have that $\mathbb{P}[B \mid \mathcal{F}_{t-1}] \ge \zeta > 0$ for infinitely many t. We showed in the proof of Corollary 1 that $\mathbb{P}[A \mid \mathcal{H}_t, B] \ge \delta \varepsilon^3$. Then, under this set-up, we have $\mathbb{P}[A \mid \mathcal{F}_{t-1}] \ge \zeta \delta \varepsilon^3$ for infinitely many t.

Proof.

$$\mathbb{P}[A \mid \mathcal{F}_{t-1}] = \mathbb{E}[\mathbb{I}_A \mid \mathcal{F}_{t-1}] = \mathbb{E}[\mathbb{E}[\mathbb{I}_A \mid \mathcal{F}_{t-1}, B] \mid \mathcal{F}_{t-1}] = \mathbb{E}[\mathbb{P}[A \mid \mathcal{F}_{t-1}, B] \mid \mathcal{F}_{t-1}] = \mathbb{E}[\mathbb{E}[\mathbb{P}[A \mid \mathcal{H}_{t}] \mid \mathcal{F}_{t-1}, B] \mid \mathcal{F}_{t-1}] \\
\geq \mathbb{E}[\mathbb{E}[\mathbb{P}[A \mid \mathcal{H}_{t}, B] \mathbb{I}_B \mid \mathcal{F}_{t-1}, B] \mid \mathcal{F}_{t-1}] = \mathbb{E}[\mathbb{E}[\mathbb{P}[A \mid \mathcal{H}_{t}, B] \mid \mathcal{F}_{t-1}, B] \mathbb{I}_B \mid \mathcal{F}_{t-1}] \\
\geq \mathbb{E}[\mathbb{E}[\delta \varepsilon^3 \mid \mathcal{F}_{t-1}, B] \mathbb{I}_B \mid \mathcal{F}_{t-1}] = \mathbb{E}[\delta \varepsilon^3 \mathbb{I}_B \mid \mathcal{F}_{t-1}] = \delta \varepsilon^3 \mathbb{E}[\mathbb{I}_B \mid \mathcal{F}_{t-1}] = \delta \varepsilon^3 \mathbb{P}[B \mid \mathcal{F}_{t-1}] \\
\geq \zeta \delta \varepsilon^3 \text{ for infinitely many } t.$$

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Proof. Let \mathcal{H}_t be defined as in (??). The first step is to show that whenever $\max_i w_t^{(i)} \geq (1+\delta)/N$, $\mathbb{P}\{c_N(t) > 2/N^2 | \mathcal{H}_t\} = \mathbb{P}\{c_N(t) \neq 0 | \mathcal{H}_t\} \geq \delta \varepsilon^3$. For this purpose we need consider only weight vectors such that $w_t^{(i)} \in (0, 2/N)$ for all i; otherwise $\mathbb{P}\{c_N(t) \neq 0 | \mathcal{H}_t\} = 1$ by the definition of stochastic rounding.

for all i; otherwise $\mathbb{P}\{c_N(t) \neq 0 | \mathcal{H}_t\} = 1$ by the definition of stochastic rounding. Denote $\mathcal{S}_{N-1}^{\delta} = \{w^{(1:N)} \in \mathcal{S}_{N-1} : \forall i, 0 < w^{(i)} < 2/N; \max_i w^{(i)} \geq (1+\delta)/N\}$ for any $\delta \in (0,1)$, where \mathcal{S}_k denotes the k-dimensional simplex. Fix arbitrary $w_t^{(1:N)} \in \mathcal{S}_{N-1}^{\delta}$. Set $i^* = \arg\max_i w_t^{(i)}$ and denote $\mathcal{I} = \{i \in \{1,\ldots,N\} : w^{(i)} > 1/N\}$. Since all weights are in (0,2/N), for $i \in \mathcal{I}, \nu_t^{(i)} \in \{1,2\}$ and for $i \notin \mathcal{I}, \nu_t^{(i)} \in \{0,1\}$; and since the offspring counts must sum to N, we can write

$$\mathbb{P}\{c_N(t) \leq 2/N^2 | \mathcal{H}_t\} = \mathbb{P}(\nu_t^{(i)} = 1 \,\forall i \in \{1, \dots, N\} | \mathcal{H}_t) \\
= \mathbb{P}(\nu_t^{(i)} = 1 \,\forall i \in \mathcal{I} | \mathcal{H}_t) \\
= \prod_{i \in \mathcal{I}} \mathbb{P}(\nu_t^{(i)} = 1 | \nu_t^{(j)} = 1 \,\forall j \in \mathcal{I} : j < i; \mathcal{H}_t) \\
= \mathbb{P}(\nu_t^{(i^*)} = 1 | \mathcal{H}_t) \prod_{\substack{i \in \mathcal{I} \\ i \neq i^*}} \mathbb{P}(\nu_t^{(i)} = 1 | \mathcal{H}_t^{(i)} = 1; \nu_t^{(j)} = 1 \,\forall j \in \mathcal{I} : j < i; \mathcal{H}_t) \\
\leq \mathbb{P}(\nu_t^{(i^*)} = 1 | \mathcal{H}_t).$$

The final inequality holds with equality when $|\mathcal{I}| = 1$, i.e. the only weight larger than 1/N is $w_t^{(i^*)}$. Thus $\mathbb{P}\{c_N(t) > 2/N^2 | \mathcal{H}_t\}$ is minimised on $\mathcal{S}_{N-1}^{\delta}$ when only one weight is larger than 1/N, in which case the values of the other weights do not affect this probability.

Define $w_{\delta'} = \{(1,\ldots,1) + \delta' e_{i^*} - \delta' e_{j^*}\}/N$ for fixed $i^* \neq j^*$ and $\delta' \in (0,1)$, where e_i denotes the *i*th canonical basis vector in \mathbb{R}^N . Using the lower bound (??) on $p_1(w_t^{(i)}) = \mathbb{P}(\nu_t^{(i)} = \lfloor Nw_t^{(i)} \rfloor + 1 \mid \mathcal{H}_t)$,

$$\mathbb{P}\{c_N(t) \le 2/N^2 \mid \mathcal{H}_t, w_t^{(1:N)} = w_{\delta'}\} = \mathbb{P}(\nu_t^{(i^*)} = 1 \mid \mathcal{H}_t, w_t^{(1:N)} = w_{\delta'})$$
$$= 1 - p_1\{(1 + \delta')/N\} \le 1 - \delta' \varepsilon^3.$$

We conclude that $\mathbb{P}\{c_N(t) > 2/N^2 | \mathcal{H}_t, \max_i w_t^{(i)} \ge (1+\delta)/N\} \ge \min_{\delta' > \delta} \delta' \varepsilon^3 = \delta \varepsilon^3$.

A slight modification of this argument yields $\mathbb{P}\{c_N(t)>2/N^2|\mathcal{H}_t,\min_i w_t^{(i)}\leq (1-\delta)/N\}\geq \delta\varepsilon^3$. Whenever $\max_i w_t^{(i)}-\min_i w_t^{(i)}\geq 2\delta/N$, either $\max_i w_t^{(i)}\geq (1+\delta)/N$ or $\min_i w_t^{(i)}\leq (1-\delta)/N$, so we have $\mathbb{P}\{c_N(t)>2/N^2|\mathcal{H}_t,\max_i w_t^{(i)}-\min_i w_t^{(i)}\geq 2\delta/N\}\geq \delta\varepsilon^3$. Thus $\mathbb{P}\{c_N(t)>2/N^2|\mathcal{H}_t\}\geq \delta\varepsilon^3\mathbb{I}_{\max_i w_t^{(i)}-\min_i w_t^{(i)}\geq 2\delta/N}$. Using the D-separation established in Appendix ?? combined with the tower property, we have

$$\mathbb{P}\{c_N(t) > 2/N^2 \mid \mathcal{F}_{t-1}\} = \mathbb{E}_t[\mathbb{P}\{c_N(t) > 2/N^2 \mid \mathcal{H}_t\}] \ge \delta \varepsilon^3 \mathbb{P}(\max_i w_t^{(i)} - \min_i w_t^{(i)} \ge 2\delta/N \mid \mathcal{F}_{t-1}),$$

which is bounded below by $\zeta \delta \varepsilon^3$ for infinitely many t. Hence,

$$\sum_{t=0}^{\infty} \mathbb{P}\{c_N(t) > 2/N^2 \mid \mathcal{F}_{t-1}\} = \infty.$$

By a filtered version of the second Borel–Cantelli lemma (see for example Durrett, 2019, Theorem 4.3.4), this implies that $c_N(t) > 2/N^2$ for infinitely many t, almost surely. This ensures, for all $t < \infty$, that $\mathbb{P}\{\exists s < \infty : \sum_{r=1}^s c_N(r) \ge t\} = 1$, which by definition of $\tau_N(t)$ is equivalent to $\mathbb{P}\{\tau_N(t) = \infty\} = 0$.

References

Durrett, R. (2019), *Probability: Theory and Examples*, Cambridge Series in Statistical and Probabilistic Mathematics, 5 edn, Cambridge University Press.