

Randomised roundings

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Definition 1. Let $X > 0$. A random variable $Y : \mathbb{R}_+ \rightarrow \mathbb{N}$ is a *randomised rounding* of X if Y takes the values

$$Y = \begin{cases} \lfloor X \rfloor & \text{with probability } 1 - X + \lfloor X \rfloor \\ \lfloor X \rfloor + 1 & \text{with probability } X - \lfloor X \rfloor \end{cases}$$

Note that we therefore have $\mathbb{E}[Y] = X$.

Definition 1 generalises easily to multivariate X . A randomised rounding of a vector $X \in \mathbb{R}_+^N$ produces a vector $Y \in \mathbb{N}^N$ such that $\mathbb{E}[Y] = X$. Y can thus be used to construct an unbiased resampling scheme.

Such popular resampling schemes as systematic resampling and residual resampling with stratified residuals are based on randomised roundings. In each scheme, the marginal distributions of the offspring counts are as in Definition 1, where $X = Nw_t^{(i)}$ and $Y = v_t^{(i)}$. The variety in rounding-based schemes arises from the interactions between these individual offspring counts.

In this document we will consider a general resampling scheme based on randomised rounding, and compare its genealogical properties to those of multinomial resampling. Let (w_1, \dots, w_N) be a normalised weight vector (that is, $w_i \geq 0$ and $\sum w_i = 1$), let $N \geq 2$ be an integer denoting the number of particles, and let (v_1, \dots, v_N) denote the resampled offspring numbers. We will attach a superscript m or r to denote offspring numbers corresponding to multinomial and rounding-based resampling respectively.

Lemma 1. For any $i \in \{1, \dots, N\}$,

$$\mathbb{E}[(v_i)_2^{(m)} | w_i] \geq \mathbb{E}[(v_i)_2^{(r)} | w_i].$$

Proof. Using properties of the Multinomial distribution (Mosimann, 1962), we have

$$\mathbb{E}[(v_i)_2^{(m)} | w_i] = N(N-1)w_i^2$$

Directly from Definition 1, we calculate the corresponding quantity in the case of randomised rounding to be

$$\begin{aligned} \mathbb{E}[(v_i)_2^{(r)} | w_i] &= \lfloor Nw_i \rfloor (\lfloor Nw_i \rfloor - 1)(1 - Nw_i + \lfloor Nw_i \rfloor) + (\lfloor Nw_i \rfloor + 1) \lfloor Nw_i \rfloor (Nw_i - \lfloor Nw_i \rfloor) \\ &= \lfloor Nw_i \rfloor (2(Nw_i - \lfloor Nw_i \rfloor) + \lfloor Nw_i \rfloor - 1) \end{aligned}$$

We define the difference

$$\begin{aligned} \Delta_i &:= \mathbb{E}[(v_i)_2^{(m)} | w_i] - \mathbb{E}[(v_i)_2^{(r)} | w_i] \\ &= N^2 w_i^2 - Nw_i^2 - 2\lfloor Nw_i \rfloor (Nw_i - \lfloor Nw_i \rfloor) - \lfloor Nw_i \rfloor^2 + \lfloor Nw_i \rfloor \\ &= N^2 w_i^2 + \lfloor Nw_i \rfloor^2 - 2Nw_i \lfloor Nw_i \rfloor - Nw_i^2 + \lfloor Nw_i \rfloor \\ &= (Nw_i - \lfloor Nw_i \rfloor)^2 - Nw_i^2 + \lfloor Nw_i \rfloor \end{aligned}$$

Then we have to show that $\Delta_i \geq 0$ for all $N \geq 2$ and $w_i \in [0, 1]$. Consider the following cases.

Case $w \in [0, 1/N)$.

Then $\lfloor Nw_i \rfloor = 0$, so we have

$$\Delta_i = Nw_i - Nw_i^2 + 0 = Nw_i(1 - w_i) \geq 0.$$

Case $w \in ((N-1)/N, 1]$.

Then $\lfloor Nw_i \rfloor = N-1$, and we have

$$\Delta_i = (Nw_i - (N-1))^2 - Nw_i^2 + N - 1 = \dots = N(N-1)(w_i - 1)^2 \geq 0.$$

Case $w = k/N$ for some $k \in \{1, \dots, N-1\}$.

Then $\lfloor Nw_i \rfloor = Nw_i = k$ and we have

$$\Delta_i = 0 - k^2/N + k = k(1 - k/N) \geq 0.$$

Case $w \in (k/N, (k+1)/N)$ for some $k \in \{1, \dots, N-2\}$

Then $\lfloor Nw_i \rfloor = k$, and

$$\Delta_i = (Nw_i - k)^2 - Nw_i^2 + k = \dots = N(N-1)(w_i - k/(N-1))^2 - \frac{k^2N}{N-1} + k^2 + k.$$

Discarding the positive first term, we have $\Delta \geq -\frac{k^2N}{N-1} + k^2 + k = k(1 - k/(N-1)) \geq 0$.

For each N , any $w_i \in [0, 1]$ falls into one of the above cases, so we conclude that $\Delta_i \geq 0$ for all N, w_i . \square

References

Mosimann, J. E. (1962), ‘On the compound multinomial distribution, the multivariate β -distribution, and correlations among proportions’, *Biometrika* **49**(1/2), 65–82.