

CORRECTION

# Correction: A Likelihood Approach to Estimate the Number of Co-Infections

Kristan A. Schneider, Ananias A. Escalante

The authors wish to acknowledge an error that was overseen in the proof of Result 1 in [1]. By carefully inspecting the authors' proof, it becomes clear that an MLE cannot exist if  $N_k = N$  for at least one  $k$ . This is easily seen from the authors' formula for  $\hat{p}_k$  in Result 1, where  $N_k = N$  implies  $\hat{p}_k = 1$ , a contradiction since then  $L(\lambda, \hat{\mathbf{p}}|\mathbf{x}) = -\infty$ . All results however remain valid as stated if their Assumption 1 is replaced by the following version.

**Assumption 1** Assume that the sum over the lineages' prevalences is larger than one, but no alleles is 100% prevalent. In other words, more than one lineage is found in at least one infection, i.e.,  $\sum_{k=1}^n N_k > N$  and all lineages are not found in every infection, i.e.,  $N_k \neq N$  for all  $k$ .

By replacing Assumption 1 with the version above in [1], the results hold without modifications. All other modifications that need to be made in the article are minor and obvious. However, the case  $N_k = N$  for at least one  $k$  was not properly addressed. This occurred because it was overseen that the proof of in Result 1 is not applicable then. What goes wrong in this case? The answer is somewhat subtle. Heuristically, this contradiction occurs because no point in the parameter space is a critical point, i.e., a point at which all derivatives of  $L$  vanish. However, for any fixed  $\lambda$ ,  $L(\lambda, \mathbf{p}|\mathbf{x})$  attains a maximum for some  $\hat{p}^{(\lambda)}$ , with  $0 < \hat{p}_k^{(\lambda)} < 1$ . The reason is that  $L(\lambda, \mathbf{p}|\mathbf{x}) = -\infty$  for  $\mathbf{p} \in \text{bd}\mathcal{S}_n$  (where  $\mathcal{S}_n$  denotes the  $n-1$ -dimensional simplex). For  $\lambda \rightarrow 0$ ,  $L(\lambda, \hat{\mathbf{p}}^{(\lambda)}|\mathbf{x}) \rightarrow -\infty$ . Hence,  $L(\lambda, \hat{\mathbf{p}}^{(\lambda)}|\mathbf{x})$  is necessarily monotonically increasing in  $\lambda$ , implying that no MLE exists. In mathematical terms this can be formulated as follows:

**Remark 1** Assume that at least one lineage is found in every sample, i.e.,  $N_k = N$  for at least one  $k$ , but not all are found in every sample, i.e.,  $N_k \neq N$  for at least one  $j$ . Then, the log-likelihood function does not attain a maximum. However, its smallest upper bound is



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$$\sup_{p \in \mathcal{S}_n, \lambda > 0} L(\lambda, \mathbf{p}|\mathbf{x}) = \sum_{\substack{k=1 \\ N_k \neq N}}^n (N - N_k) \log\left(1 - \frac{N_k}{N}\right).$$

The supremum is reached in the limit of any sequence  $(\lambda_t, \mathbf{p}_t)$  with  $\lim_{t \rightarrow \infty} \lambda_t = \infty$ ,  $\lim_{t \rightarrow \infty} p_k^{(t)} = -\log\left(1 - \frac{N_k}{N}\right)$  if  $N_k \neq N$  and  $\lim_{t \rightarrow \infty} p_k^{(t)} \lambda_t = \infty$  if  $N_k = N$ .

**Proof.** Because  $L(\lambda, \mathbf{p}|\mathbf{x})$  is bounded by 0, the supremum exists. Furthermore, a sequence  $(\lambda_t, \mathbf{p}_t)$  exists with  $L(\lambda_t, \mathbf{p}_t|\mathbf{x}) \rightarrow \sup_{p \in \mathcal{S}_n, \lambda > 0} L(\lambda, \mathbf{p}|\mathbf{x})$ .

Without loss of generality let  $N_1, \dots, N_m < N$  and  $N_{m+1} = \dots = N_n = N$ . Hence,

$$\begin{aligned}
 L(\lambda, \mathbf{p} | \mathbf{x}) &= -N \log(e^\lambda - 1) + \sum_{k=1}^n N_k \log(e^{\lambda p_k} - 1) \\
 &= N \log \frac{(e^{\lambda p_{m+1}} - 1) \cdot \dots \cdot (e^{\lambda p_n} - 1)}{e^\lambda - 1} + \sum_{k=1}^m N_k \log(e^{\lambda p_k} - 1) \\
 &= N \log \frac{(1 - e^{-\lambda p_{m+1}}) \cdot \dots \cdot (1 - e^{-\lambda p_n})}{1 - e^{-\lambda}} e^{-\lambda(1 - p_{m+1} - \dots - p_n)} + \sum_{k=1}^m N_k \log(e^{\lambda p_k} - 1) \\
 &= N \log \frac{(1 - e^{-\lambda p_{m+1}}) \cdot \dots \cdot (1 - e^{-\lambda p_n})}{1 - e^{-\lambda}} e^{-\lambda(p_1 + \dots + p_m)} + \sum_{k=1}^m N_k \log(e^{\lambda p_k} - 1).
 \end{aligned} \tag{1}$$

Let  $(\lambda_t)$  be any monotone sequence with  $\lim_{t \rightarrow \infty} \lambda_t = \infty$ . Moreover, let  $c_k > 0$  for  $k = 1, \dots, m$ . Now let  $\mathbf{p}_t$  be a sequence satisfying  $\lim_{t \rightarrow \infty} p_k^{(t)} \lambda_t = c_k$  for  $k = 1, \dots, m$  and  $\lim_{t \rightarrow \infty} p_k^{(t)} \lambda_t = \infty$  for  $k = m + 1, \dots, n$ . Without loss of generality let  $p_k^{(t)} = \frac{c_k}{\lambda_t}$  for  $k = 1, \dots, m$  and  $p_k^{(t)} = \frac{1}{n-m} (1 - \sum_{k=1}^m \frac{c_k}{\lambda_t})$  for  $k = m + 1, \dots, n$ . For sufficiently large  $t$  this sequence is defined and  $\mathbf{p}_t \in \mathcal{S}_n$ . Hence,

$$\begin{aligned}
 \lim_{t \rightarrow \infty} L(\lambda_t, \mathbf{p}_t | \mathbf{x}) &= N \log 1 \cdot e^{-c_1 - \dots - c_m} + \sum_{k=1}^m N_k \log(e^{c_k} - 1) \\
 &= -N(c_1 + \dots + c_m) + \sum_{k=1}^m N_k \log(e^{c_k} - 1).
 \end{aligned}$$

Next define  $f(c_1, \dots, c_m) := \lim_{t \rightarrow \infty} L(\lambda_t, \mathbf{p}_t | \mathbf{x})$ . Note that this definition is independent of the sequence  $(\lambda_t, \mathbf{p}_t)$ , with  $\lambda_t \mathbf{p}_t \rightarrow (c_1, \dots, c_m, \infty, \dots, \infty)$  for  $t \rightarrow \infty$ .

The next aim, is to identify potential maxima of  $f$ . Clearly,  $\frac{\partial f}{\partial c_k} = -N + N_k \frac{e^{c_k}}{e^{c_k} - 1}$ . Equating the partial derivatives to zero gives  $\hat{c}_k = -\log(1 - \frac{N_k}{N})$ . The Hessian matrix is given by  $H = -\text{diag}(N_k \frac{e^{c_k}}{(e^{c_k} - 1)^2})_{k=1, \dots, m}$  and clearly negative definite. Thus,  $f$  attains a global maximum at  $\hat{c}_k = -\log(1 - \frac{N_k}{N})$ . Therefore  $f(\hat{c}_1, \dots, \hat{c}_m) \leq \sup_{\mathbf{p} \in \mathcal{S}_n, \lambda > 0} L(\lambda, \mathbf{p} | \mathbf{x})$ .

If  $(\lambda_t, \mathbf{p}_t)$  is any sequence with  $\lambda_t p_k^{(t)} \rightarrow \infty$  for a  $k$  with  $1 \leq k \leq m$ , it is easily seen from (1) that  $\lim_{t \rightarrow \infty} L(\lambda_t, \mathbf{p}_t | \mathbf{x}) = -\infty$ . Moreover, if  $\lambda_t p_k^{(t)} \rightarrow c_k < \infty$  for  $1 \leq k \leq m$  and at least one  $k$  with  $m + 1 \leq k \leq n$ , without loss of generality  $\lambda_t p_k^{(t)} \rightarrow c_k < \infty$  for  $m + 1 \leq k \leq n$ , (1) implies

$$\begin{aligned}
 \lim_{t \rightarrow \infty} L(\lambda_t, \mathbf{p}_t | \mathbf{x}) &= -N(c_1 + \dots + c_m) + \sum_{k=1}^m N_k \log(e^{c_k} - 1) + N \sum_{k=m+1}^n \log(1 - e^{-c_k}) \\
 &< -N(c_1 + \dots + c_m) + \sum_{k=1}^m N_k \log(e^{c_k} - 1)
 \end{aligned}$$

implying that this limit is less than the maximum of  $f$ . The above considerations imply that the supremum of the log-likelihood function must be the maximum of  $f$ . Deriving  $f(\hat{c}_1, \dots, \hat{c}_m)$  finishes the proof.

The case that  $N_k = N$  for all  $k$  is treated in [1]. Moreover, obviously in Remark 1 of [1] a misprint occurred. The expression  $\sum_{k=1}^n N_k \geq N$  needs to be replaced by  $\sum_{k=1}^n N_k > N$ , while the same expression needs to be replaced by  $\sum_{k=1}^n N_k = N$  in the paragraph below Result 2.

## Reference

1. Schneider KA, Escalante AA (2014) A Likelihood Approach to Estimate the Number of Co-Infections. PLoS ONE 9(7): e97899. <https://doi.org/10.1371/journal.pone.0097899> PMID: 24988302