## General comments

The clarity of the methods used in the paper is much improved. I have one substantial comment.

## Specific comments

On page 4 lines 2-4 you wrote: "and is normalized to 1. It defines the probability density function of the environment, given no prior information and some observed count  $y_{k0}$  of taxon k."

I wondered in my comment (2 January 2017) whether this normalization was needed and concluded that it was not. This conclusion disregarded the somewhat ad-hoc combination in equation (6). Without the normalization of  $L_y$  and  $L_p$  (but with normalization of prob(x)), we still have the same result for  $\eta = 0$  and  $\eta = 1$  and the same path of solutions, but for the meaning of  $\eta$  would change. In particular, the calculations for  $\eta = 0.5$  would give different results (the current result can be obtained with a different value for  $\eta$ ).

The authors can either keep the current normalization (but please deleted the "It defines.." sentence, as it adds nothing) and the ad-hoc combination or change things to a perhaps more defendable combination as follows.

Two models are proposed to infer about *x*. The first model (M<sub>1</sub>) says that abundance percentages relate to *x* and the second model (M<sub>2</sub>) that the presences relates to *x*. So, first a posterior is made on the basis of the first model, say  $prob_y(x)$ ), and then one on the basis of the second model, say  $prob_p(x)$ . Both probability densities are normalized, of course. Let the prior probability of M<sub>1</sub> be  $\eta$ . Then the final posterior of *x* is

$$prob(x) = \eta \ prob_{\nu}(x) + (1 - \eta) prob_{\nu}(x).$$

This construct gives again the same path of solutions as the one with early normalization, but the new construct is a little bit more logical. If desired so, it even allows estimation of  $\eta$  on the basis of the posterior probability of M<sub>1</sub>.

I note for clarity (and you may wish to add it) that equation (4) follows from the law of total probability:

$$prob(y_{k0}|x) = \sum_{j} prob(y_{k0}, SRC_{jk}|x) = \sum_{j} prob(SRC_{jk}) prob(y_{k0}|SRC_{jk}, x)$$

Typo: Equation number (3) is italic.

Cajo ter Braak, Wageningen , 3 January 2017