

1 **Significant phylogenetic signal is not enough to trust phylogenetic predictions**

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9 Arising from A. Cantwell-Jones et al. *Nature Plants* <https://doi.org/10.1038/s41477-022-01100-6> (2022)

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11 **Abstract**

12 In a recent study, Cantwell-Jones et al. (2022) proposed a list of 1044 species as
13 promising key sources of B vitamins based primarily on phylogenetic predictions. To
14 identify candidate plants, they fitted lambda models of evolution to edible species with
15 known values in each of six B vitamins (232 to 280 species) and used the estimated
16 parameters to predict B-vitamin profiles of edible plants lacking nutritional data (6460
17 to 6508 species). The latter species were defined as potential sources of a B vitamin if
18 the predicted vitamin content was $\geq 15\%$ towards recommended dietary allowances for
19 active females between 31-50 years per 100 g of fresh edible plant material consumed.
20 Unfortunately, the reliability of the predictions that informed the list of candidate
21 species is questionable due to insufficient phylogenetic signal in the data (Pagel's λ
22 between 0.171 and 0.665) and a high incidence of species with missing values (over
23 95% of all the species analyzed in the study). We found that of the 1044 species
24 proposed as promising B-vitamin sources, 626 to 993 species showed accuracies that
25 were indistinguishable from those obtained under a white noise model of evolution (i.e.
26 random predictions conducted in absence of any phylogenetic structure) in at least one

27 of the vitamins, which proves the weakness of the inference drawn from imputed
28 information in the original study. We hope this commentary serves as a cautionary note
29 for future phylogenetic imputation exercises to carefully assess whether the data meet
30 the requirements for the predictions to be valuable, or at least more accurate than
31 expected by chance.

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33 **Main**

34 In a recent study, Cantwell-Jones et al.¹ proposed a list of 1044 species as promising
35 key sources of B vitamins based primarily on phylogenetic predictions. To identify
36 candidate plants, they fitted lambda models of evolution to edible species with known
37 values in each of six B vitamins and used the estimated parameters to predict B-vitamin
38 profiles of edible plants lacking nutritional data. The latter species were defined as
39 ‘sources’ of a B vitamin if the predicted vitamin content was $\geq 15\%$ towards
40 recommended dietary allowances for active females between 31–50 years per 100 g of
41 fresh edible plant material consumed. This phylogenetic imputation exercise is exciting
42 and inspiring, more so as it would have the potential to be replicated for further goals
43 such as, for example, identifying species of pharmaceutical interest. Unfortunately, the
44 predictions from this study are questionable. The predictive capability of lambda
45 models is primarily determined by the amount of phylogenetic signal in the known data
46 (i.e. the extent to which closely related species share similar values in the trait of
47 interest), which means that predictions based on low phylogenetic signals, even if
48 statistically significant, are typically valueless²⁻⁴. Further, simulations have shown that
49 even under ‘strong’ phylogenetic signal (i.e. Pagel’s $\lambda = 1$), acceptable predictions can
50 only be expected when the most recent common ancestor (MRCA) of a target species
51 and its closest relative with known trait value (hereafter ‘predictive MRCA’) is

52 relatively recent⁴. As such, the older the predictive MRCAs, the lesser the difference
53 between phylogenetic imputations and random predictions (Fig. 1). Based on 232 to 280
54 nutritionally known species (observed data), Cantwell-Jones et al.¹ predicted B-vitamin
55 profiles for 6460 to 6508 nutritionally unknown species (over 95% of all the species
56 analyzed in the study), and they did so relying on very weak-to-moderate phylogenetic
57 signals in the observed data (Pagel's λ between 0.171 and 0.665). These figures suggest
58 that the predictions conducted by the authors are unreliable due to both insufficient
59 phylogenetic signal and a high incidence of relatively old predictive MRCAs (Fig. 1).
60 Moreover, the unreliability of their predictions calls into question the proposed list of
61 1044 species as promising sources of B vitamins.

62 Cantwell-Jones et al.¹ conducted leave-one-species-out cross-validation trials on
63 the observed values of the B vitamins they analyzed (they referred to this procedure as
64 “jackknifing”) and estimated 95% confidence intervals (CI) around each re-estimated
65 (predicted) value. They found that $\geq 91.4\%$ of nutritionally known species had measured
66 (observed) values within the 95% CI of the values predicted in the trials. Here, we
67 conducted the same analysis but randomizing the authors' dataset in a two-step
68 procedure. First, the observed values of each vitamin were reshuffled across the species
69 with known value in the vitamins 500 times, and then each reshuffled set ($n = 500$ sets
70 per vitamin) was checked to show complete lack of phylogenetic signal (i.e. $\lambda < 0.001$
71 and $p > 0.05$). Otherwise, the values were reshuffled iteratively until the conditions
72 were met. We found that $\geq 92.8\%$ of the observed values in the randomized sets sit
73 within 95% CIs of their corresponding predicted values (Supplementary Data 1-5). This
74 demonstrates that the authors simply found the null expectation of the analysis, which
75 deems the CI-based evidence of their study as invalid.

76 On the other hand, the authors used generalised least squares models (with a
77 variance structure in the error term) to test the strength of the relationship between
78 predicted and observed values, and they found significant relationships for all the
79 vitamins. Here, we used a simple and more intuitive prediction coefficient (P^2) to assess
80 the overall predictive power of the lambda models they employed:

$$81 \quad P^2 = 1 - \frac{\sum_{i=1}^n (\hat{y}_i - y_i)^2 / n}{s_y^2}$$

82 where \hat{y}_i and y_i are respectively the predicted and observed value for the nutritionally
83 known species i and s_y^2 is the sample variance of the observed values of the trait^{4,6}. P^2
84 can be interpreted similarly as Ezekiel's adjusted coefficient of determination⁷, so that
85 $P^2 = 1$ when all predictions perfectly match the observations and $P^2 = 0$ when the model
86 is no better in predicting values than simply taking the mean of the observed values
87 (note that P^2 has no negative boundary, as there is no theoretical limit to how badly a
88 model can predict observed trait values). We found that P^2 was close to zero in all cases
89 except for folate, which showed a slightly higher (yet weak) score (Table 1). This is not
90 surprising, as previous simulations have shown that phylogenetic predictions perform
91 very poorly when λ is lower than $\sim 0.6^4$. There are numerous studies that caution against
92 using phylogenies alone to predict missing data under low phylogenetic signal^{2-4,8,9}, yet
93 none of them was acknowledged in Cantwell-Jones et al.¹

94 In line with the authors' observation that "median differences between predicted
95 and observed values for each nutrient were <33% of the standard deviation across
96 species", it could be argued that some of the predictions may still be useful despite the
97 discouraging results from the leave-one-species-out cross-validation trials (Table 1).
98 Thus, we used the method described in Molina-Venegas et al.⁴ to assess the expected
99 species-level accuracy of the predictions that informed the list of candidate plants (see
100 Supplementary Information for details). Depending on whether p -values were

101 Bonferroni-corrected ($n = 1616$, two-sample Wilcoxon tests) and the nominal alpha
102 criterion (5% or 0.1%), we found that of the 1044 species proposed as promising B-
103 vitamin sources, 626 to 993 species showed accuracies that were indistinguishable from
104 those obtained under a ‘white noise’ model of evolution (i.e. random predictions
105 conducted in absence of any phylogenetic structure) in at least one of the vitamins
106 (Supplementary Data 6). Moreover, regarding the species that showed statistically
107 significant accuracies, it should not be concluded that their predictions are necessarily
108 valuable, only that they are better than drawing values at random independently of the
109 phylogeny. This is illustrated by the rather modest maximum accuracy that was
110 recorded across all candidate species and B vitamins analyzed, which corresponded to
111 *Hordeum bulbosum* and niacin with only 54.4% of P^2_{sim} values ≥ 0.75 (Supplementary
112 Data 7). These results are not surprising either. The extremely high incidence of missing
113 values in the dataset makes predictive MRCAs prone to be relatively old (Fig. 1), which
114 has a proven negative impact on prediction accuracy^{4,8}. As such, simulations show that
115 only species whose predictive MRCA is younger than ~10% of the height of the
116 phylogeny are expected to show consistently accurate predictions (i.e. at least 75% of
117 P^2_{sim} values ≥ 0.75) under a scenario of ‘strong’ phylogenetic signal (i.e. $\lambda = 1$)⁴.

118 It is very exciting to see a burgeoning interest in connecting phylogenetic
119 information with human well-being, but researchers should be clear on the limitations of
120 phylogenetic predictive methods and utilize imputed information with caution and
121 restraint, especially if the goal is employing individual predictions separately (e.g. to
122 evaluate if a species has potential to be ‘source’ of a B vitamin, as in Cantwell-Jones et
123 al.¹). The dataset gathered by Cantwell-Jones et al.¹ is impressive, and it would not be
124 surprising if the 1044 species they proposed as candidate sources attract the attention of
125 professionals willing to invest resources in measuring their B-vitamin content.

126 Unfortunately, the reliability of the predictions that informed the list of candidate
127 species ranges between weak to very weak. While the authors did not intend to predict
128 the exact nutrient content of species but to evaluate if the predicted values were above
129 or below a certain threshold, the fact that most of their predictions are indistinguishable
130 from pure randomness proves the weakness of the inference drawn from imputed
131 information in their study. We hope this commentary serves as a cautionary note for
132 future phylogenetic imputation exercises to carefully assess whether the data meet the
133 requirements for the predictions to be valuable, or at least more accurate than expected
134 by chance.

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143 **Author contributions**

144 R.M.-V. conceived and discussed the idea with M.A.R. and I.M.-C., performed the
145 calculations and led the writing, I.M.-C. drew the figure, and all the authors contributed
146 to the writing.

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148 **Competing Interests**

149 The authors declare no competing interests

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151 **Additional information**

152 **Supplementary information.** The online version contains supplementary material

153 available at XXX

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155 **Data availability.** All the datasets generated for this study are available from the

156 Figshare Digital Repository at <https://doi.org/10.6084/m9.figshare.19682880.v1>.

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160 **Table 1.** Phylogenetic signal λ of each B vitamin analyzed across N nutritionally known

161 species and prediction coefficients P^2 derived from the leave-one-species-out cross-

162 validation trials.

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B vitamin	N	λ	P^2
Thiamine	280	0.293	-0.010
Riboflavin	277	0.192	0.035
Niacin	278	0.665	0.086
Pantothenic acid	232	0.171	0.023
Folate	256	0.302	0.183

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165 **Figure 1.** Conceptual model on the expected accuracy of phylogenetic predictions as a

166 function of phylogenetic signal and distance to predictive MRCAs, this is, the MRCA

167 of each target species and its closest relative with known trait value. Accuracy is

168 expected to be acceptable only under strong phylogenetic signal and relatively recent

169 predictive MRCAs (short distances), and predictions will be uncertain if any of these

170 conditions are not met. The probability of finding relatively old predictive MRCAs

171 (long distances) increases with the amount of missing data, which may lead to the
172 worst-case scenario if phylogenetic signal is weak.

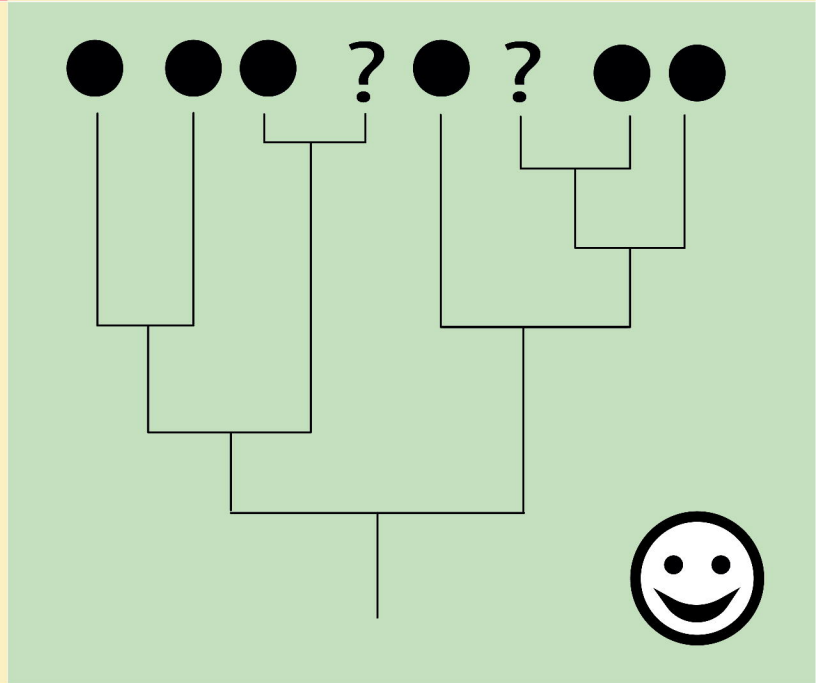
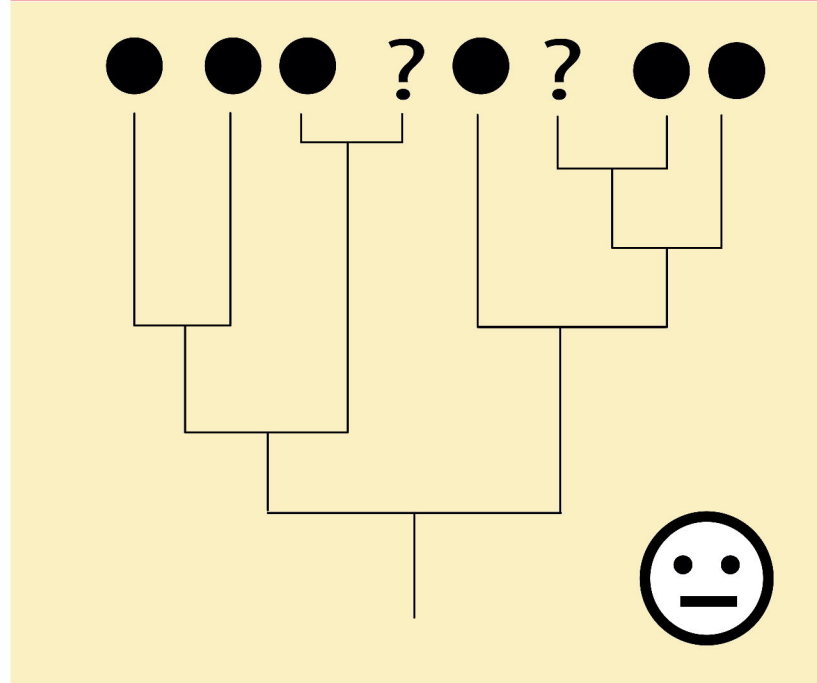
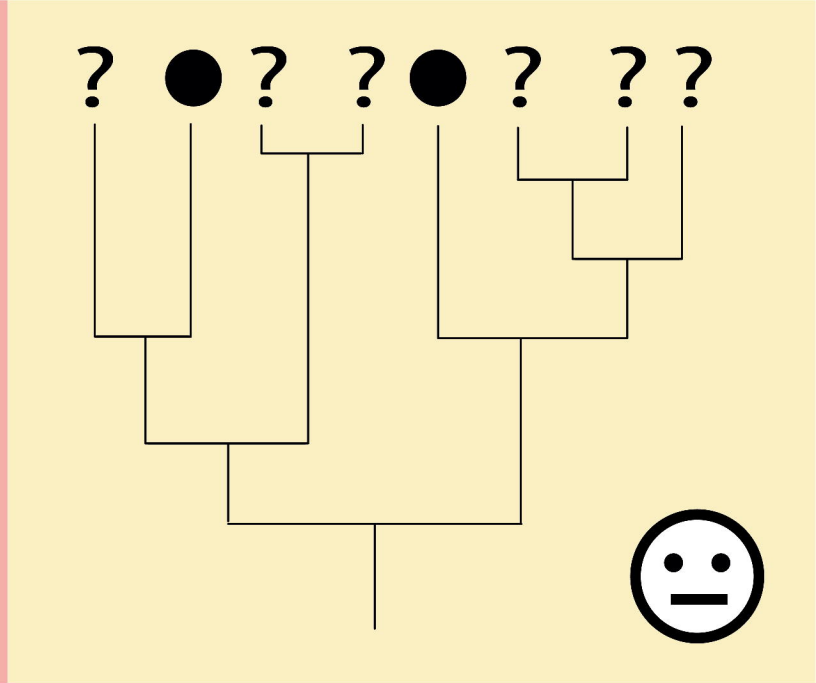
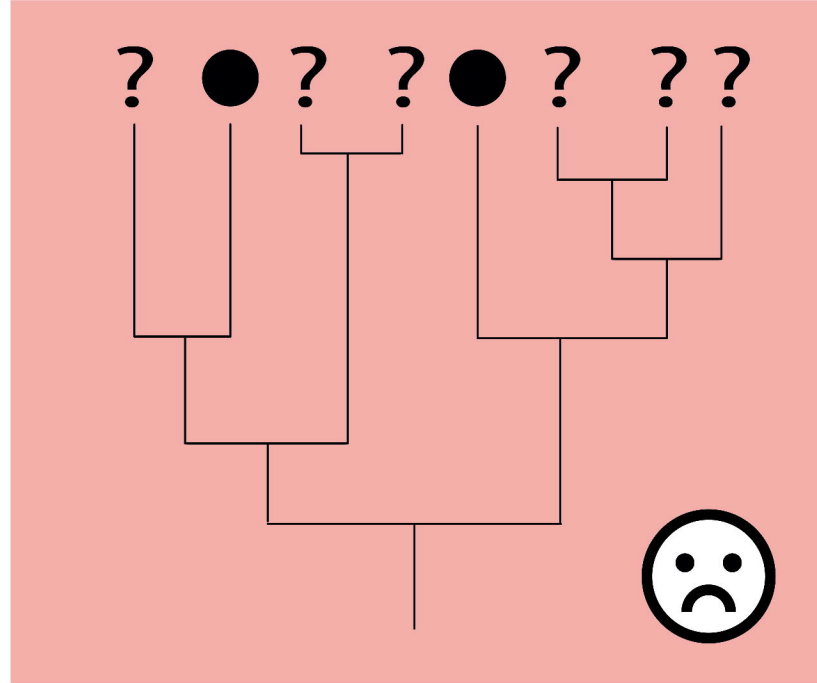
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Distance to predictive MRCA

Longer ↑
Shorter ↓



← Weaker Stronger →

Phylogenetic signal