

## Additional file 2 — The choice of $Z_k$ in the constraint

While testing ERaBLE we have realised that setting  $Z_k = 1$  or  $Z_k = N_k$ , for all  $k \in \{1, 2, \dots, m\}$  can cause important overestimations of the scale factors  $\hat{\alpha}_k$  for genes only present in a small group of closely related taxa. This phenomenon is strictly linked to the strong underestimation of a minority of gene rates — and the slight overestimation of the majority of gene rates — observed for SDM-based methods, which also use the constraint with  $Z_k = 1$ . In our experiments we have set  $Z_k = N_k \sum_{i,j \in L_k} \delta_{ij}^{(k)}$ , which largely solves this problem, despite being rather heuristic. In this additional file, we show the importance of the constraint used by ERaBLE with a very simple example.

We construct a small data set consisting of just two nucleotide alignments, those of exons ENSG00000066654\_THUMPDI\_000 and ENSG00000127423\_AUNIP\_000 obtained from OrthoMaM after trimAl filtering. We call them  $G_1$  and  $G_2$ , respectively. For simplicity we only keep the sequences of six species, those in the set  $L = \{Gorilla, Homo, Pan, Bos, Erinaceus, Sorex\}$ . Since  $G_1$  is only sampled in primates, we have  $L_1 = \{Gorilla, Homo, Pan\}$ , and  $L_2 = L$ . Alignment lengths are  $N_1 = 489$  for  $G_1$ , and  $N_2 = 855$  for  $G_2$ . Figure 7 shows a phylogenetic tree for these data.

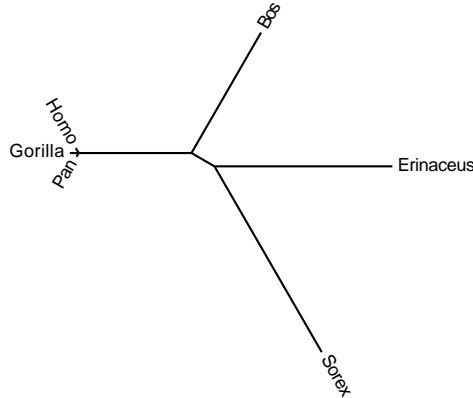


Figure 7 – Maximum likelihood tree (PhyML with model TN93+ $\Gamma_8$ ) obtained on the concatenation of the two alignments in the example.

Fig. 8 shows the distance matrices estimated for  $G_1$  and  $G_2$  (left column), and the different behaviours of ERaBLE with  $Z_k = 1$  and  $Z_k = N_k \sum_{i,j \in L_k} \delta_{ij}^{(k)}$  (middle and right column, respectively). The behaviour for  $Z_k = N_k$  is similar to that for  $Z_k = 1$ , and not shown here for brevity. A quick comparison of  $\Delta_1$  and  $\Delta_2$  suggests that the rate of  $G_1$  is higher than that of  $G_2$  (note that, in two cases out of three,  $\delta_{ij}^{(1)}$  is more than the double than  $\delta_{ij}^{(2)}$ ). We then expect that  $\hat{\alpha}_1 < \hat{\alpha}_2$ . However, when  $Z_k = 1$ , the opposite happens: solving problem (6) leads to  $\hat{\alpha}_1 = 1.73$  and  $\hat{\alpha}_2 = 0.274$ .

The key observation to understand why this happens is that  $\Delta_1$  only contains very closely related species (all great apes, see also Fig. 7), so its distances are very small relative to many of those in  $\Delta_2$ , which can be up to about 30 times larger. As a consequence, if  $\hat{\alpha}_1 \leq \hat{\alpha}_2$  the value of the objective function  $Q(\hat{\alpha}, \hat{b})$  is dominated by the differences  $\hat{\alpha}_2 \delta_{ij}^{(2)} - \hat{d}_{ij}$ , that is, the difference between  $\hat{\alpha}_2 \Delta_2$  and  $\hat{D} = (\hat{d}_{ij})$ . It is then intuitive that a way to reduce  $Q(\hat{\alpha}, \hat{b})$  is to simultaneously reduce the scale of  $\hat{\alpha}_2 \Delta_2$  and  $\hat{D} = (\hat{d}_{ij})$ , which can be achieved by decreasing the value of  $\hat{\alpha}_2$  (and consequently increasing that of  $\hat{\alpha}_1$ , given that for  $Z_k = 1$  their mean is constrained to be 1).

This is precisely what is happening when setting  $Z_k = 1$  in our example (middle column in Fig. 8): instead of having  $\hat{\alpha}_1 < \hat{\alpha}_2$ , ERaBLE produces a small  $\hat{\alpha}_2 = 0.274$  and a large  $\hat{\alpha}_1 = 1.73$ . Compared to the alternative constraint (right column), where we have  $\hat{\alpha}_1 < \hat{\alpha}_2$  as expected, it is clear that this results

Input distances						Trivial constraint $Z_k = 1$						Our constraint $Z_k = N_k \sum_{i,j \in L_k} \delta_{ij}^{(k)}$					
						$\hat{\alpha}_1$	$\hat{\alpha}_2$					$\hat{\alpha}_1$	$\hat{\alpha}_2$				
						1.73	.274					0.538	1.00				
	Homo	Pan	Bos	Erinac.	Sorex	Homo	Pan	Bos	Erinac.	Sorex	Homo	Pan	Bos	Erinac.	Sorex		
Gorilla	.0203	.0135	$\Delta_1$			Gorilla	.0351	.0233	$\hat{\alpha}_1 \Delta_1$			Gorilla	.0109	.0073	$\hat{\alpha}_1 \Delta_1$		
Homo	.0148					Homo	.0256					Homo	.0080				
Gorilla	.0087	.0099	.278	.354	.432	Gorilla	.0024	.0028	.0763	.0970	.118	Gorilla	.0087	.0099	.278	.354	.432
Homo	.0062		.279	.342	.432	Homo	.0017		.0766	.0937	.118	Homo	.0062		.279	.342	.432
Pan			.286	.345	.429	Pan			.0783	.0945	.118	Pan			.286	.345	.429
Bos			.419		.422	Bos			.115		.116	Bos			.419		.422
Erinac.	$\Delta_2$		.446			Erinac.	$\hat{\alpha}_2 \Delta_2$		.122			Erinac.	$\hat{\alpha}_2 \Delta_2$		.446		
Gorilla						Gorilla	.0127	.0117	.0777	.0985	.116	Gorilla	.0087	.0099	.284	.360	.424
Homo						Homo	.0104		.0774	.0983	.116	Homo	.0069		.280	.357	.420
Pan						Pan			.0764	.0973	.115	Pan			.282	.358	.422
Bos						Bos			.107		.124	Bos			.390		.454
Erinac.						Erinac.	$\hat{D}$					Erinac.	$\hat{D}$				

Figure 8 – **Changing behaviour of ERaBLE with different constraints.** In this example, ERaBLE is run on the two distance matrices  $\Delta_1$  and  $\Delta_2$  on the left. Setting  $Z_k = 1$  results in the  $\hat{\alpha}_k$  values and matrices  $\hat{\alpha}_1 \Delta_1$ ,  $\hat{\alpha}_2 \Delta_2$  and  $\hat{D} = (\hat{d}_{ij})$  in the middle column. Our chosen setting for  $Z_k$  results in more reasonable values for these quantities (right column), as explained in the text.

in significantly smaller differences  $\hat{\alpha}_2 \delta_{ij}^{(2)} - \hat{d}_{ij}$  for most distances, the only exceptions being the three distances between primates. Note that the scale of  $\hat{\alpha}_2 \Delta_2$  cannot be reduced indefinitely, as then the scale of  $\hat{\alpha}_1 \Delta_1$  becomes too large, and the fit of  $\hat{D}$  with the distances between primates in the two rescaled distance matrices becomes too loose (note that for  $Z_k = 1$  the distances between primates in  $\hat{\alpha}_1 \Delta_1$  and  $\hat{\alpha}_2 \Delta_2$  already differ by an order of magnitude and the fit with  $\hat{D}$  is very poor).

The, admittedly heuristic, approach that we have adopted in our experiments, that is, setting  $Z_k = N_k \sum_{i,j \in L_k} \delta_{ij}^{(k)}$ , essentially prevents the genes only appearing in few and closely related taxa from having an influence on the constraint. Thus, it is the  $\hat{\alpha}_k$  for the remaining genes that are constrained to have a weighted average of 1 (where the weight depends on the length of their sequence, as in equation (5)). As a consequence, these  $\hat{\alpha}_k$  cannot be reduced together with  $\hat{D}$ , as we showed for  $Z_k = 1$ . In our example, the new constraint is  $24 \cdot \hat{\alpha}_1 + 3838 \cdot \hat{\alpha}_2 = 3862$ , which is roughly equivalent to imposing  $\hat{\alpha}_2 = 1$ . The results are then much more realistic than with  $Z_k = 1$ : for example the rescaled matrices  $\hat{\alpha}_1 \Delta_1$  and  $\hat{\alpha}_2 \Delta_2$  are now much closer on their common entries (right column in Fig. 8).