### 1 Title page

- Article title: Assessing the Effects Imputation on ED Values
- Running head: Assessing the Effects Imputation on ED Values
- 4 Authors: K. Bodie Weedop<sup>1</sup>, William D. Pearse<sup>1</sup>
- $^{\rm 1}$  Department of Biology & Ecology Center, Utah State University, 5305 Old Main Hill,
- 6 Logan UT, 84322
- <sup>7</sup> \*To whom correspondence should be addressed: will.pearse@usu.edu and bodie.weedop@aggiemail.usu.ed
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- , Abstract
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#### 11 Introduction

Evidence from the fossil record and present-day studies argue we are in the midst of, or entering, a sixth mass extinction (Barnosky et al. 2011; Ceballos et al. 2015), such that more 13 species than ever are declining and/or in danger of extinction across a range of environments 14 (Wake & Vredenburg 2008; Thomas et al. 2004). Habitat destruction (Brooks et al. 2002), 15 invasive species (Molnar et al. 2008), climate change (Pounds et al. 2006), and disease (Lips 16 et al. 2006) are some of the leading causes of species declines globally. Conservation biologists 17 seeks to reverse<del>overcome</del> these declines and their detrimental effects on species populations, 18 but in reality they have limited resources with which to do so.. This challenge, termed the Even still, researchers and conservationists are confronted with "Noah's Ark problem" (Weitzman 1998), is the basis for modern conservation prioritisation and triage (XXX define 21 triage) or an unfortunate reality of insufficient, finite resources to confront the increasing amount species requiring conservation effort. Conservation triage has provided an efficient decision making process for allocating finite 24 resources to obtain the greatest return (Bottrill et al. 2008). A sentence about how triage 25 requires decision-making, and so you need a metric to guide that decision process. One of 26 these triage strategies which have been introduced and used most widely is the EDGE metric 27 (Evolutionary Distinction and Globally Endangered; Isaac et al. 2007). This method pri-28 oritizes species according to two metrics: Eevolutionary EvolutioDdistinctiveness (ED) and Gglobal Eendangerment (GE). ED measures relative contributions to phylogenetic diversity made by each species within a particular clade (Isaac et al. 2007). Such contributions are as-31 sessed by quantifying the amount of branch length which is unique to each species within the 32 overall phylogeny. GE values are assessed by assigning numerical values to each of the World Conservation Union (IUCN) Red List Categories. As species become increasingly threatened and are placed into more concerning categories (e.g., e.g. from Vulnerable to Endangered), the GE numerical value increases. Increases in either ED or GE place a particular species at a higher priority for conservation effort.

In the event of missing DNA or trait data, species are often difficult or not able to be placed onto a phylogeny. Even in the face of such uncertainty and missing data, it is understandable that conservation biologists want to make prioritizations. However, if we are using a quantitative method for prioritizing species, we should remain consistent even when uncertainty arises. To our knowledge, a proper and efficient method for prioritizing species where there is missing data is still untested. This issue pertains mainly to the calculation of ED than GE. The IUCN has collected data on most major clades, and has a strategy for assigning Red Listing values these species which we know little information and are considered Data Deficient (DD). IUCN and other conservation organizations support focus on DD species just the same as Critically Endangered and Endangered species to ensure consistency (Rodrigues et al. 2006). The major area of uncertainty in phylogenetic prioritisation is phylogenetic data. In the past, missing species data and poorly resolved trees have been addressed using imputation (Collen et al. 2011; Isaac et al. 2012; Jetz et al. 2014). However, to our knowledge, there has been no systematic investigation of the efficacy of such imputation, both in 51 terms of the accuracy with which imputed ED values are estimated, and the effect on other 52 known species' scores. Indeed, it is unclear whether any significant information on ED is gained by imputing species which cannot be placed on the phylogeny. It is also not well understood how simply removing missing species, compared to performing imputation, would effect ED values. It may be that simply excluding missing species may be less intrusive than imputation. In searching for a solution for missing species, we may be negatively affecting correct ED values and disrupting EDGE rankings in the process. As the desire to use ED and phylogenies for conservation triage grows, the importance of such tests and a consensus on how to resolve cases of phylogenetic uncertainty becomes more urgent.

Here we assess the extent to which EDGE rankings based upon imputed phylogenies can be used within applied conservation biology. To do this, we use an imputation approach... Here,

we assess and compare the impact that missing species versus phylogenetic imputation has
upon ED values. In doing so, we hope to understand the effect that both methods have on ED
values and offer a viable solution for dealing with missing data species. Missing species were
simulated and removed from trees in two ways: randomly and in a phylogenetically biased
manner. Additionally, we tested how ED values were affected by resolving and imputing
polytomies of varying sizes on a phylogeny. We found that ED values are by removing
missing species. Here is a reference to figure.

### Methods

Here we use a simulation approach to test the effect of removing and imputing species on a phylogeny on subsequent ED (Evolutionary Distinctiveness) scores of species. We attempted to demonstrate the effects of removing or imputing species which cannot be placed onto a phylogeny. Since empirical studies do not (to our knowledge) impute GE (Global Endangerment) scores for species, and our focus here is on the importance of phylogenetic structure, we focus on the impact of imputing ED values. While assessing the impact of dropping species and phylogenetic imputations, we were primarily focused on ED values. In testing the effects on these values, we remain focused upon ED as a single variable. We exclude GE because it would only add complexity while not providing any additional information to our particular investigation.

In each test, we simulated 100 phylogenetic trees and manipulated each tree's tips or elade.

All simulations, calculations, and analyses were performed using R (version 3.4.0; R Core
Team 2017). For each combination of parameter values in a simulation, we performed 100
replicate simulations. Original and manipulated trees were simulated under a pure-birth
Yule model using the sim.bdtreesim.bdtree function geiger'geiger' R package (Harmon et
al. 2007). This particular model was chosen to maintain simplicity. Results from this simple

model should be applicable to other more complex scenarios. Also, to reduce uncertainty, we used the same model throughout each of the simulations. In reality, we would be estimating the parameters of the model which the phylogeny is built upon. The function ed.calced.cale within the R package caper eaper was used to calculate ED values for each tree (Orme 2013).

We assessed the impact that removing missing species has upon ED values using the correlation of all ED values for the tips remaining within both trees. To evaluate the effect that imputation has upon ED values, we calculated ED for all tips in both the original and manipulated trees while excluding the focal clade where imputation has occurred. These ED values were compared using a correlation. Additionally, we did the same calculations and comparison using only the original focal clade and its' simulated replacement. If missing species have no effect upon ED values, we expect a high, positive correlation coefficient between the original tree and its' manipulated counterpart.

#### Assessing the impact of missing species on EDGE-listing

Our first set of simulations assesses the impact that species missing from a phylogeny have 101 on estimated ED scores. If, when species are missing from a tree, the ED scores of the 102 remaining species in the tree are XXX this implies XXX. If missing species has a negative 103 effect on ED values, then the correlation between the ED of the species in the original tree 104 and the same tree with a fraction of tips removed in some manner should be significantly 105 different from 1. We performed simulations on phylogenies of different sizes (number of taxa: 106 XXX, XXX), removing constant fractions of tips from the tree (0%, 1%, 2%, ..., 19%, 20%). 107 To investigate the degree of this effect, we removed tips from simulated trees (Number of 108 taxa = 64, 128, 256, 512, 1024, 2048, 4096) at random and by phylogenetic clusters. To 109 assess the effect under varying amounts of uncertainty, fractions of tips dropped ranged from 110

#### 0 to 0.2 of total tree tips.

Missing species at random was simulated by selecting species at random without replacing, 112 and removing those species from the tree. This randomization had no regard for phyloge-113 netic structure. Missing species related by some character trait was tested by simulating 114 character trait values for each tip. These simulations were all performed under a constant 115 rate Brownian-motion model ( $\sigma^2=0.5$  par = 0.5, starting root value = 1). Tips were dropped if their character trait values place them into the upper quantile which had been selected to 117 be dropped. More specifically, if the fraction to be dropped was 0.1, species within the 90th 118 quantile of character trait values are dropped. This is equivalent to Felsenstein's threshold 119 model (Felsenstein & Felenstein 2004) state why this model is a useful one—re-state the 120 propoerty it has, linking it back to why this is a useful set of simulations to be doing. 121

#### Assessing the impact of phylogenetic imputations

We tested the impact of imputing missing species onto a clade of a particular size (sizes 3, 123 4, 5, ..., 30, 31, 323 through 32) which originated from a tree of a particular size (Nnumber 124 of taxa = 64, 128, 256, 512, 1024). To simulate the effect that phylogenetic imputation 125 has upon these simulated trees, we randomly chose clades within each tree and treated it 126 as a polytomy to be resolved. The clade selected was removed from the original tree and 127 a new separate tree of the same size was simulated under the pure-birth model used before 128 and placed back where the original clade was removed. Thus we have imputed each clade 129 under the same model used to generate it. In an empirical study, this would be done by... 130 and so our method is being generous because... By doing this, we replicated the process of 131 imputation of a clade which has been resolved. To ensure that this is representative of cases 132 where imputation is used, 100 repetitions of this simulation were performed across different 133 parameter settings.

To assess whether clades, once imputed, had similar ED scores, we ... We also looked at ranks, because... We statistically modelled these as a function of ..., hypothesising that each would matter because...

#### Results

While the random loss of species from a phylogeny does not appear to affect the ED values
of the remaining species, phylogenetically-patterned loss does (Fig. 1 and Table ??). Under
both random and phylogenetically patterned loss, XXX increases with XXX, although the
effect is much more extreme (XXX times; Table ??) for XXX.ED values for remaining species
were significantly affected by the fraction of species which were removed (Table 1). However,
different effects are seen when dropping species at random and in clustered manner (Fig.
1). Dropping species at random has a reduced effect when compared to the effect which
dropping species in a clustered manner has on remaining ED values (Fig. 1).

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.9991	0.0003	3752.33	0.0000
Fraction of Species Dropped	-0.3434	0.0021	-164.89	0.0000
Random Treatment	-0.0004	0.0004	-0.96	0.3360
Number of Species Overall	0.0000	0.0000	0.13	0.8932
Fraction of Species Dropped:Random Treatment	0.3129	0.0029	106.25	0.0000
Random Treatment:Number of Species Overall	0.0000	0.0000	9.20	0.0000

Table 1: ANCOVA model summary describing the effect of dropping species on remaining species ED Values. The fraction of species dropped significantly affects the the remaining ED values. Dropping the fraction at random had a reduced compared to dropping species in a clustered manner ( $F_{29688,5} = 12090, R^2 = 0.6706, p<0.0001$ ).

We find no support for a correlation between the imputed and true ED values for a species within an imputed clade (Fig. 2, table XXX). We do find evidence that, when imputing larger clades, the variation in the correlation is lesser (quantile regression), but this could be due to XXX. ED values for the full tree while excluding the focal clade remain at 1 and

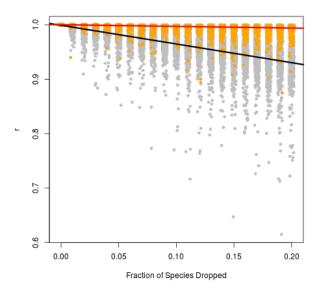


Figure 1: R-values plotted against the fraction of species dropped at random versus clustered manner. The color of data points denote whether species were dropped at random (orange; n = 100) or in clustered manner (grey; n = 100). The regression lines are demonstrating the relationship when species are dropped at random (black) and in a clustered manner (red).

unaffected. However, ED values for the imputed clades are significantly affected by the use 151 of imputation. As the size of the focal clade increases, the informative value of the ED values 152 within the clade decreases (Fig. 2). However, even when imputing smaller clades, ED values 153 did not regularly reflect the true ED values (Table 2). We foundOur analysis demonstrates 154 that measures of the true<del>original</del> phylogeny such as phylogentic diversity (PD), lambda, 155 Colless' Index, skew, and kurtosis do not provide any indication that imputation would 156 negatively affect ED values (Appendix A). Additionally, Just as imputed ED values did not 157 reflect true ED values, the rankings of species within the focal clade were altered significantly 158 under imputation (Fig. 3; Table XXX). Our model suggests that with increases in the size of 159 the imputed clade and overall number of species, species within the clade are ranked farther 160 from their true ranking (Table 3). For example More specifically, our model suggests that 161 by imputing a clade of three species within a phylogeny of 128 species, the species within 162 the clade would be 60 rankings from their true rank on average. ED values outside of the focal clade were not affected by imputation. While ED values within the focal clades were affected exclusively by imputation, a notable error rate in ranking crucial species correctly is present (Appendix B).

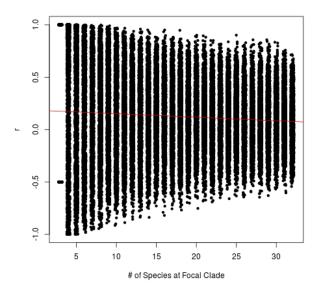


Figure 2: R-values plotted against the number of species at focal clade. Each data point denotes a correlative comparison between ED values within the focal clades where imputation has occurred. The regression line (red) and trend even closer to zero demonstrates the decrease in informative value of the imputed ED values. This is reinforced by the visual narrowing of r-values around zero.

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	0.1852	0.0533	3.47	0.0005
Size of Focal Clade	-0.0034	0.0002	-16.56	0.0000
Size of Phylogeny	-0.0001	0.0001	-0.41	0.6855
PD	0.0001	0.0001	0.37	0.7108
Lambda	-0.0012	0.0524	-0.02	0.9812
Colless' Index	0.0016	0.0022	0.72	0.4687
Skew	0.0043	0.0088	0.48	0.6288
Kurtosis	-0.0005	0.0009	-0.63	0.5269

Table 2: Effect of Clade Size on Imputed ED Values. The intercept describes that the correlation between the true and imputed values begins quite low. As the clade size increases, this correlation only tends toward zero. The total number of species in the full phylogeny along with measures of the true phylogenetic diversity, lambda, Colless' Index, skew, and kurtosis show no significant effect.  $(F_{47992,7} = 39.57, R^2 = 0.006, p < 0.0001)$ .

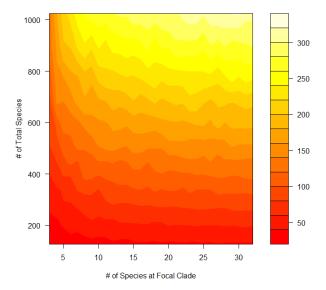


Figure 3: Mean ranking error of species within the focal clade. The gradient on the right demonstrates average number of posistions within the full ranking that focal clade species shifted from their true rank. While controlling for the size of the full phylogeny and focal clade, species within the focal clade were, on average, ranked far from the true rank.

	Estimate	Std. Error	t value	$\Pr(> t )$
(Intercept)	-1.6344	0.0332	-49.29	0.0001
Size of Focal Clade	0.0900	0.0010	91.22	0.0001
Size of Phylogeny	0.5179	0.0013	383.99	0.0001

Table 3: Effect of Clade Size and Total Species on Ranking Error. Model demonstrating the relationship between focal clade species ranking error and the size of imputed clade and overall phylogeny. Square-root transformations have been applied to both ranking error and size of phylogeny. Significant increases ranking error are seen when increasing sizes of both the imputed clade and phylogeny ( $F_{47997,2}=77890,\,R^2=0.7644,\,p<0.0001$ ).

167 Discussion

168 Acknowledgments

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# 222 A. Effect of Measures of the True, Full Phylogenies

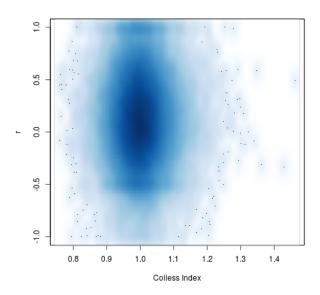


Figure 4: Effect of the True Colless Index of FullPhylogeny.

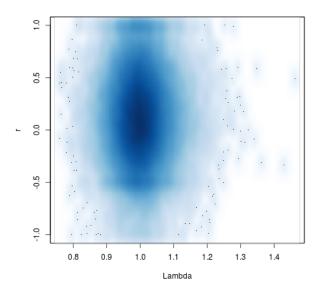


Figure 5: Effect of the True Lambda of Full Phylogeny.

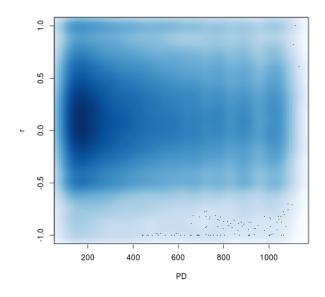


Figure 6: Effect of True PD of Full Phylogeny.

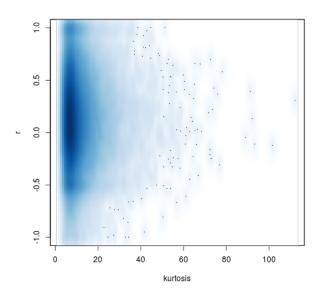


Figure 7: Effect of the True Kurtosis of Full Phylogeny.

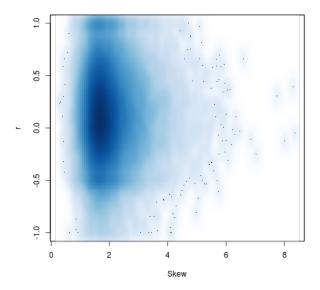


Figure 8: Effect of the True Skew of Full Phylogeny.

# 223 B. Error Rate in Top Rankings

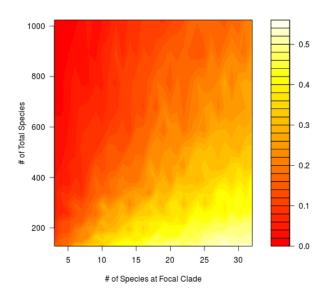


Figure 9: Mean error rate in the ranking of top 50 species.

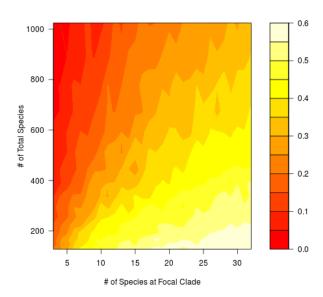


Figure 10: Mean error rate in the ranking of top 100 species.

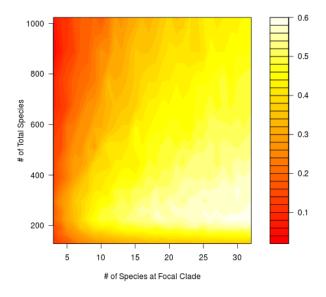


Figure 11: Mean error rate in the ranking of top 200 species.

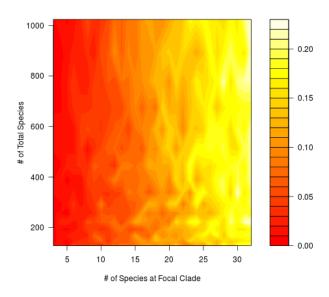


Figure 12: Mean error rate in the ranking of top 5% of species.

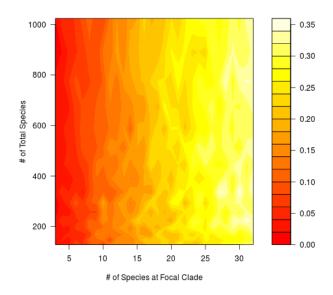


Figure 13: Mean error rate in the ranking of top 10% of species.

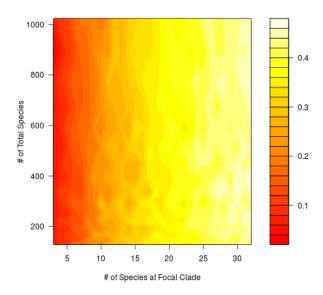


Figure 14: Mean error rate in the ranking of top 20% of species.