

1 **Reducing the bias due to unknown relationships in measuring the steepness of a dominance**  
2 **hierarchy**

3

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11 **Abstract**

12 Measuring the steepness of a dominance hierarchy is important for classifying a social system in  
13 a continuum between egalitarian and despotic. For this, often the steepness-slope from de Vries  
14 and colleagues (*Animal Behaviour*, 2006, **71**, 585-592) is used. It compares the cardinal and  
15 ordinal dominance rank of each individual using the slope of the linear regression. The  
16 disadvantage of this measure is that the slope becomes lower the higher the proportion of unknown  
17 relationships (dyads without interactions). In the present paper, we investigate what causes this  
18 bias, and propose a solution. We show: (1) that the bias is due to the treatment of unknown  
19 relationships by the dominance index currently used in this methodology, the David's score  
20 (namely by assuming, among other things, an equal number of wins and losses for both members  
21 of the pair). (2) Instead, using the Average Dominance Index (the average proportion of wins by  
22 each individual from all its opponents) reduces the bias due to unknown relationships, because it

23 excludes unknown relationships, and (3) the standard error of the steepness slope based on the  
24 Average Dominance Index is smaller. (4) The two indices (David's score and Average Dominance  
25 Index) result in similar steepness-slopes when all relationships are known. For comparing the two  
26 indices we use empirical data (from four group-years of wild vervet monkeys) and data from a  
27 computational model on dominance interactions in a group (DomWorld). We conclude that the  
28 Average Dominance Index (compared to the David's score) is preferable for measuring the  
29 steepness-slope.

30 Keywords: Steepness of hierarchy; David's score; Average Dominance Index; Dominance  
31 interactions; Egalitarian society; Despotic society

32

33 In social animals, the steepness of a group's dominance hierarchy classifies its dominance style  
34 along a continuum from "egalitarian" at one end to "despotic" at the other (e.g., Vehrencamp,  
35 1983; van Schaik, 1989). Conceptually, the hierarchy's steepness represents the difference in  
36 ability to win agonistic interactions between individuals of adjacent rank: when the difference is  
37 large, the hierarchy is steep and the society is despotic, and when it is small, the hierarchy is  
38 shallow and the society egalitarian (Vehrencamp, 1983). The most widely used measure of  
39 steepness of the hierarchy is that proposed by De Vries et al. (2006), with over 300 citations to  
40 date (in Web of Science Core Collection). De Vries et al. (2006) quantified steepness as the  
41 absolute value of the slope of the (least squares straight) line describing the relationship between  
42 the group member's ordinal rank (on the X axis) and its cardinal rank (on the Y axis). The ordinal  
43 rank of an individual represents its relative position in the hierarchy (e.g., first, second ...  $n^{\text{th}}$   
44 position, with  $n$  equal the number of individuals), whereas its cardinal rank represents a continuous  
45 measure of its competitive ability calculated using a dominance index. As dominance index, de

46 Vries et al. (2016) used the David's Score, normalized for group size. This index of steepness is  
47 problematic, however, because its value decreases with the percentage of pairs of group members  
48 (dyads) without competitive interactions, the so-called 'unknown relationships' (Klass & Cords,  
49 2011). In empirical datasets unknown relationships are common. For example, a meta-analysis of  
50 101 interaction matrices from 55 published studies found unknown relationships to comprise, on  
51 average, 26% of the total number of dyads (standard deviation = 23%), reaching values up to 70%  
52 in few cases (Shizuka & McDonald, 2012). Therefore, a steepness measure that is less influenced  
53 by unknown relationships is desirable.

54 In the present work, we propose a modification of the steepness-slope that is less influenced by  
55 unknown relationships. We argue that the sensitivity of the steepness-slope to unknown  
56 relationships is caused by the David's Score's treatment of unknown relationships as being  
57 perfectly egalitarian, i.e., as if the two individuals had the same number of wins and losses (de  
58 Vries, Stevens, & Vervaecke, 2006). We minimize this sensitivity by excluding unknown  
59 relationships from the calculation of the dominance index as is done in the Average Dominance  
60 Index, ADI (Hemelrijk, Wantia, & Gyax, 2005). The Average Dominance Index ranks  
61 individuals according to their average proportion of winning from all their opponents (excluding  
62 group members with whom they did not fight). Thus, we expect the steepness-slope to be less  
63 affected by unknown relationships when we use the Average Dominance Index in the measure. A  
64 steepness measure that is influenced by unknown relationships less is preferable in empirical  
65 datasets, because here unknown relationships are common.

66 We first explain the mathematical formula of the David's Score and why it is particularly  
67 susceptible to unknown relationships in the group. Next, we explain the formula of the Average  
68 Dominance Index and show that it is less impaired by these.

69 **The David's Score**

70 The David's Score consists of four elements, the first two increasing the final score, while the last  
71 two decrease it, with higher scores representing better competitive ability, thus higher dominance  
72 (de Vries et al., 2006):

73 
$$\text{David's score} = W + W_{\text{weighted}} - L - L_{\text{weighted}}$$

74 The calculation of the four elements for an individual  $i$  is as follows: the first element,  $W$ , is the  
75 sum of dyadic proportions of fights won by the individual  $i$  with all other group individuals  $j, j \neq$   
76  $i, j \leq N$ , where  $N$  is the group size.

77 
$$W_i = \sum_{j=1}^{N(j \neq i)} \frac{\text{wins}_{ij}}{\text{n of fights}_{ij}} \quad (\text{equation 1})$$

78 The second element,  $W_{\text{weighted}}$ , is the sum of the dyadic proportion of fights won by the individual  
79  $i$  with each other individual  $j, j \neq i$ , weighted by the combat capabilities of each interacting  
80 partner, represented by the interacting partner's  $W$  value, as in equation 1.

81 
$$W_{i \text{ weighted}} = \sum_{j=1}^{N(j \neq i)} W_j \frac{\text{wins}_{ij}}{\text{n of fights}_{ij}} \quad (\text{equation 2})$$

82 The third element,  $L$ , is the sum of the dyadic proportion of fights lost by individual  $i$  with each  
83 other individual  $j, j \neq i$ . The proportion of losses for each dyad is complementary to the proportion  
84 of victories.

85 
$$L_i = \sum_{j=1}^{N(j \neq i)} \frac{\text{losses}_{ij}}{\text{n of fights}_{ij}} \quad (\text{equation 3})$$

86 The fourth element,  $L_{\text{weighted}}$ , is the sum of the dyadic proportion of fights lost by the individual  
 87  $i$  with each other individual  $j, j \neq i$ , weighted by the propensity of the interacting partner to lose a  
 88 fight, represented by the partner's  $L$  value.

$$89 \quad L_{i \text{ weighted}} = \sum_{j=1}^{N(j \neq i)} L_j \frac{\text{losses}_{ij}}{\text{n of fights}_{ij}} \quad (\text{equation 4})$$

90 The David's Score for each individual is the sum of the first two elements (equation 1 and 2) minus  
 91 the last two (equation 3 and 4). When there are no interactions between two individuals,  $i$  and  $j$ ,  
 92 ( $\text{n of fights}_{ij} = 0$ ), the four parts of the David's Score for that dyad<sub>ij</sub> are zero. This is also true in  
 93 a perfectly egalitarian relationship, i.e., where the two individuals  $i$  and  $j$  have (1) the same number  
 94 of wins ( $\text{wins}_{ij}$ ) and losses ( $\text{losses}_{ij}$ ) from each other and (2) the same number of total wins ( $W_i$   
 95 and  $W_j$ , equation 1) as losses ( $L_i$  and  $L_j$ , equation 4) with all group members ( $W_i=W_j=L_i=L_j$ ). In  
 96 this theoretical case, the first two elements of the David's Score would be equal to the second two  
 97 and the sum would be zero for that dyad<sub>ij</sub>. This zero value is true for the David's Score of both  
 98 individuals,  $i$  and  $j$ .

99 For example, we can take two interaction matrices: one in which individuals A and B have the  
 100 same number of wins and losses with each other and with two other individuals, C and D  
 101 (interaction matrix I, Table 1), and another one in which individuals A and B never interact with  
 102 each other, and have non-egalitarian relationships (i.e., different number of wins and losses) with  
 103 C and D (interaction matrix II, Table 1). In the egalitarian case, the contribution of the single dyad<sub>ab</sub>  
 104 to the David's Score of individual A is zero ( $0.5 + 0.25 - 0.5 - 0.25 = 0$ , equations 1 to 4). The  
 105 same is true for the David's Score of individual B ( $0.5 + 0.25 - 0.5 - 0.25 = 0$ ). In the second  
 106 matrix, the contribution of the single dyad<sub>ab</sub> to the David's Score of individual A is also zero ( $0 +$

107  $0 - 0 - 0 = 0$ ). The same is true for the David's Score of individual B ( $0 + 0 - 0 - 0 = 0$ ). The  
 108 effect of  $dyad_{ab}$  in the calculation of the David's Score of both individuals (A and B) is the same  
 109 in both hierarchies. This example shows that even if the two individuals (A and B) have different  
 110 competitive abilities (A is higher ranking than B in the second interaction matrix), a possible  
 111 unknown relationship between the two is treated as if they had the same number of wins and losses  
 112 in their dyad ( $dyad_{ab}$ ) as well as with the other individuals in the hierarchy ( $dyad_{ac}$ ,  $dyad_{ad}$ ,  $dyad_{cb}$ ,  
 113  $dyad_{db}$ ). This result is illogical and highlights further how the treatment of unknown relationships  
 114 in the David's Score is a methodological limitation.

115 So far, the values of the David's Score range between  $-N(N-1)/2$  to  $+N(N-1)/2$ , where  $N$  is the  
 116 group size, and this causes the steepness-slope to range between 0 and  $N$ . To make the values of  
 117 the steepness-slope range between 0 and 1, de Vries et al. (2006) normalized the David's Score,  
 118 such that its values range between 0 and  $N-1$  (equation 5). The maximum value of steepness is  
 119 calculated for a perfectly linear dominance hierarchy in which all individuals consistently defeat  
 120 others ranking below them, and lose all interactions from others above them (de Vries et al., 2006).

121 
$$\text{normDS} = \frac{DS + N(N - 1)/2}{N} \quad (\text{equation 5})$$

122 **The Average Dominance Index**

123 The Average Dominance Index, ADI (Hemelrijk et al., 2005), is calculated for each individual  $i$   
 124 as its average proportion of winning from each opponent  $j$  with whom it had an interaction  
 125 (equation 6).

126 
$$ADI_i = \frac{W_i}{n \text{ of opponents}_{ij}(*)} \quad (\text{equation 6})$$

127 (\*) opponents<sub>*ij*</sub> with at least one interaction

128 When a relationship between two individuals *i* and *j* is unknown (n of fights<sub>*ij*</sub> = 0), that partner  
129 does not contribute to  $W_i$  nor to the n of opponents<sub>*ij*</sub>, and therefore is omitted from the calculation  
130 of its average proportion of winning, ADI<sub>*i*</sub>.

131 When used to calculate a steepness-slope that should range between 0 and 1, the Average  
132 Dominance Index needs to be normalised with group size, because its un-normalised values range  
133 between 0 and 1, regardless of group size. Without normalising the Average Dominance Index,  
134 when the index is used in the steepness-slope, the values of the steepness-slope become smaller  
135 the larger the group size, because the maximal cardinal rank equals the group size (N = group size)  
136 while the maximal Average Dominance Index is independent of group size and remains 1.  
137 Therefore, to get a steepness-slope between 0 and 1, the Average Dominance Index needs to be  
138 normalised with group size. When multiplying the Average Dominance Index by N-1 (the group  
139 size minus one), the normalised Average Dominance Index ranges between 0 to N-1, the same  
140 range as that of the normalized David's Score and consequently, the steepness values using the  
141 Average Dominance Index range between 0 and 1 too. The range of the values of the steepness-  
142 slope is the same when calculated with the normalized David's Score and normalized Average  
143 Dominance Index. From now on (apart from in tables, in abbreviations, and in the Conclusion  
144 section, in which the normalization is expressly mentioned) when referring to these dominance  
145 indices, we imply their normalized version.

#### 146 **Empirical data**

147 Our empirical data concern aggressive interactions of vervet monkeys (*Chlorocebus pygerythrus*).  
148 collected at the Inkawu Vervet Project (IVP) in the Mawana game reserve, South Africa, between

149 2011 and 2018. Three groups of habituated wild vervet monkeys took part in the study: ‘Ankhase’  
150 (AK), ‘Kubu’ (KB), and ‘Noha’ (NH). Habituation began in 2010 in AK and NH, and in 2013 in  
151 KB. Data were collected by several trained observers. All IVP observers are trained to identify  
152 each monkey by individual bodily and facial features (eye-rings, scars, color, shape etc.) and have  
153 to pass an identification test for all the individuals of a group. Before beginning the collection of  
154 behavioural data, observers had to pass an inter-observer reliability test with Cohen’s kappa > .80  
155 for each data category between two observers. Data were initially collected on handheld computers  
156 (Palm Zire 22) using Pendragon software version 5.1 and, from the end of August 2017, on tablets  
157 (Vodacom Smart Tab 2) and smartphones (Runbo F1) equipped with the Pendragon version 8.

158 The aggressive data were collected through *ad libitum* (Altmann, 1974) sampling because conflicts  
159 happened infrequently. In our matrices, we noted only conflicts that were dyadic, between adults,  
160 and had a clear outcome: the last behaviour of the winner was aggressive (namely, stare, chase,  
161 attack, hit, bite, take place) and the last behaviour of the loser was submissive (namely, retreat,  
162 flee, leave, avoid, jump aside).

163 Ethics guidelines: Our study adhered to the “Guidelines for the use of animals in research” of  
164 Association for Study of Animal Behavior and was approved by the relevant local authority,  
165 Ezemvelo KZN Wildlife, South Africa.

### 166 **Monte Carlo simulations**

167 We started by selecting four group-years with the smallest percentage of unknown relationships in  
168 the interaction matrices (Table 1). We monitored the performance of the steepness-slope using  
169 both indices (the David’s Score and the Average Dominance Index) in each group-year, starting  
170 with the original percentage of unknown relationships and progressively increasing the percentage



171 of unknown relationships. For this, we iteratively excluded one dyad at a time from the interaction  
172 matrix using Monte Carlo simulations. This is the same approach used by Klass and Cords, (2011).  
173 After the exclusion of each dyad, we calculated the steepness-slope using both indices. For each  
174 group-year, we calculated the median value of hierarchical steepness for both indices for 100 runs  
175 of the Monte Carlo simulation with each run using a new, random sequence of excluding dyads.  
176 We omitted runs in which, after the exclusion of the last dyad, one or more individuals were  
177 completely removed from the interaction matrix because they did not have any remaining  
178 interactions, because their absence would change the group size. For each Monte Carlo run, we  
179 calculated the Kendall Rank correlation between hierarchical steepness and proportion of unknown  
180 relationships. Applying the Wilcoxon signed rank test, we investigated whether the distribution of  
181 these Tau values was symmetrical around 0.

182 We compared how the two indices changed depending on the percentage of unknown  
183 relationships. For this, we calculated their percentage change in steepness as the difference in  
184 steepness between that of the original social interaction matrix and the median of the steepness  
185 when the percentage of unknown relationships approached one-fourth (25%) and two-thirds  
186 (approximately 66.7%). We chose these particular percentages to match the average (26%) and  
187 highest (up to 70%) percentage of unknown relationships found in published empirical datasets  
188 (Shizuka & McDonald, 2012). For each group-year, we took the percentage of unknown  
189 relationships closest to these thresholds, and calculated the median value of hierarchical steepness  
190 over the 100 Monte Carlo runs.

### 191 **Using a computational model to create hierarchies without unknown relationships**

192 To understand the performance of the two indices in the steepness-slope in the absence of unknown  
193 relationships, we produced artificial data on dominance interactions in the computational model,

194 DomWorld (Hemelrijk, 1999). We use a computational model because, compared to empirical  
195 data, it is easier to create a large number of dominance hierarchies and social interaction matrices  
196 in which all interactions are known. In DomWorld, individuals group and interact competitively.  
197 We derive the dominance hierarchy from the social interaction matrix of winners and losers in the  
198 group. To validate the computational model, DomWorld, against the empirical data, we  
199 investigated, for both the David's Score and the Average Dominance Index, whether an increasing  
200 percentage of unknown relationships results in values of steepness-slope that resembles those in  
201 the empirical data. To compare against empirical data we used 100 runs of Monte Carlo  
202 simulations for ten hierarchies obtained from the model DomWorld and calculated the median  
203 values and percentage decrease of hierarchical steepness at one-fourth and two-third unknown  
204 relationships for each run. We omit Monte Carlo runs in which some individuals were completely  
205 excluded from the interaction matrix because all their dyads were randomly excluded, not to affect  
206 group size. This approach is the same as in empirical data. We compared both indices also for 100  
207 hierarchies in DomWorld when all relationships were known. For each matrix, we calculated the  
208 hierarchical steepness for both indices, and compared their steepness-slopes with a linear  
209 regression.

210 Monte Carlo simulations and statistical analyses were conducted in R 4.0.3 (R Core Team, 2021).  
211 For the calculation of hierarchical steepness with the David's Score, we used the package  
212 EloRating (Neumann & Kulik, 2020) with the correction for the number of interactions in a dyad  
213 from de Vries *et al.* (2006), called  $D_{ij}$ .  $D_{ij}$  is a corrected proportion of wins for each dyad that  
214 adjusts the original proportion ( $P_{ij}$ ) by subtracting a value linked with how likely the proportion of  
215 wins ( $P_{ij}$ ) is given the amount of interactions in the dyad (equation 7: de Vries et al, 2006). The  
216 probability to observe each proportion of wins ( $P_{ij}$ ) is calculated from a uniform distribution, and

217 is equal to  $1/(n_{\text{dyad}} + 1)$ , with  $n_{\text{dyad}}$  being the number of interactions in the dyad. Thus, the value  
218 that is subtracted from the original proportion of wins for the dyad ( $P_{ij}$ ) is smaller the greater  
219 number of interactions in the dyad; the value that is subtracted approaches zero when the number  
220 of interactions tends to infinite, but can never be zero.

$$221 \quad D_{ij} = P_{ij} - \{(P_{ij} - 0.5) \times \text{likelihood}[P_{ij}]\} \quad (\text{equation 7})$$

222 In our present, we use  $D_{ij}$  over the simpler  $P_{ij}$  for the proportion of wins in a dyad in calculating  
223 the David's Score, as was suggested by de Vries et al. (2006). However, Balasubramaniam et al.  
224 (2013) suspected that with an increasing percentage of unknown relationships  $D_{ij}$  might artificially  
225 decrease steepness-slopes more than  $P_{ij}$ . Thus, we first study in the computational model  
226 DomWorld whether  $D_{ij}$  and  $P_{ij}$  differ in their bias depending on the percentage of unknown  
227 relationships.

### 228 **Testing the performance of David's Score's $D_{ij}$ versus $P_{ij}$ in the steepness-slope**

229 We used a dominance hierarchy obtained from DomWorld to compare the performance of using  
230 in the David's Score the simple proportion of wins ( $P_{ij}$ ) versus the proportion of wins corrected for  
231 chance ( $D_{ij}$ ). When we used  $D_{ij}$  we found a smaller steepness-slope than when we used  $P_{ij}$ , when  
232 all relationships were known. However, the trend for the steepness-slope with an increasing  
233 percentage of unknown relationships was the same whether we used  $D_{ij}$  or  $P_{ij}$  in calculating the  
234 steepness-slope. When the percentage of unknown relationships was close to 25%, the steepness-  
235 slope decreased by 38.9% (from 0.712 to a median value of 0.435) when we used  $D_{ij}$ , compared  
236 to a decrease of 38.8% (from 0.735 to a median value of 0.450) when we used  $P_{ij}$ . When the  
237 percentage of unknown relationships was close to 66.7%, the steepness-slope decrease for both  $D_{ij}$   
238 and  $P_{ij}$  by 83.4% (for  $D_{ij}$ : from 0.712 to a median of 0.118; for  $P_{ij}$  from 0.735: to a median of

239 0.122). Thus, from here onward we calculate the David's Score using the correction for the number  
240 of interactions in a dyad ( $D_{ij}$ ), as originally proposed by de Vries et al. (2006).

### 241 **Performance of the two indices in empirical data**

242 The percentage of unknown relationships in the matrices of the four group-years of vervet  
243 monkeys from the Inkawu Vervet Project was significantly, negatively correlated with the  
244 steepness-slope calculated with the David's Score but significantly, positively when calculated  
245 with the Average Dominance Index (Table 2). The correlation was stronger with the David's Score  
246 than with the Average Dominance Index. At approximately 25% unknown relationships,  
247 steepness-slopes changed 62.1 times more due to unknown relationships when based on the  
248 David's Score than on the Average Dominance Index and at approximately 66.7% they changed  
249 3.8 times more. At one-fourth unknown relationships (~25%), on average steepness-slopes  
250 calculated with the David's Score decreased by 28.8% (min -17.6%, max -39.2%), and with the  
251 Average Dominance Index decreased by 0.02% (min -0.05%, max +1.96%). At two-thirds  
252 unknown relationships (~66.7%), the steepness-slopes calculated with the David's Score  
253 decreased on average by 78% (min -77.5%, max -80.6%), and that by the Average Dominance  
254 Index increased by 20.3% (min +16.6%, max +22.6%).

### 255 **Validating the computational model against empirical data**

256 Results in DomWorld resemble those from the empirical data on vervet monkeys. With the David's  
257 Score, the steepness-slopes decreased on average by 38.7% at one-fourth unknown relationships,  
258 and by 83.0% at two-thirds. With the Average Dominance Index, it increased by 2.1% at one-  
259 fourth unknown relationships, and by 15.7% on average at two-thirds. Therefore, when 25% of the  
260 relationships were unknown the steepness-slopes changed approximately 18.4 times more due to

261 unknown relationships when based on the David's Score than the Average Dominance Index and  
262 5.3 times more when unknown relationships were approximately 66.7%. These results are for  
263 interaction matrices of groups of ten males in DomWorld. In the model, males are more intense in  
264 their aggression than females. From here onwards the DomWorld results are based on interaction  
265 matrices of ten males. This is because results do not change when we match sex composition and  
266 group size to those in the empirical group-years (data not shown, available on request) and we  
267 want to have a single baseline for interaction matrices across all analyses, with a large group size  
268 and moderately steep hierarchies (which we get by only including the sex with the most intense  
269 aggression in the model).

#### 270 **Testing the performance of the two indices in matrices without unknown relationships**

271 In matrices where all relationships are known (there are interactions in all dyads), steepness-slopes  
272 that use the David's Score and Average Dominance Index give similar values (linear regression of  
273 the David's Score steepness on the Average Dominance Index steepness results in intercept =  
274 0.002, slope = 0.966,  $R^2 = 0.9994$ ,  $P < 0.001$ , Figure 2). The slope in this regression was slightly  
275 below one, and the steepness-slopes using the David's Score were on average 0.022 lower  
276 (standard deviation = 0.00026) than with the Average Dominance Index.

#### 277 **Testing the combination of bias and variability**

278 The quality of an estimator also depends on its variability, with a smaller variability generally  
279 preferred among predictors with the same bias. To evaluate the contribution of the variability to  
280 the total deviation, we use the root mean squared deviation (RMSD). It is the square root of the  
281 sum of the square of the bias (difference between sample mean and value to be estimated) plus the  
282 variance of the sample. A smaller value of the RMSD is preferable. We used data from the

283 computational model, DomWorld, to have initial matrices where all relationships are known to use  
284 as a reference for the unbiased value of steepness. At each step, the deviation of steepness was  
285 calculated per dominance index as the difference between the average steepness-slopes at that step  
286 and the steepness-slopes without unknown relationships. We used ten matrices obtained from  
287 DomWorld where all relationships were known and made 100 runs of the Monte Carlo simulation  
288 for each matrix. In DomWorld, the average value of RMSD at 25% unknown relationships was  
289 0.27 for the David's Score and 0.05 for the Average Dominance Index. At 66.7%, the RMSD was  
290 0.58 for the David's Score and 0.16 for the Average Dominance Index. Therefore, steepness-slopes  
291 based on the Average Dominance Index outperform those based on the David's Score by 5.4 times  
292 at one-fourth unknown relationships and by 3.6 times at two-thirds.

## 293 **Discussion**

294 The steepness-slope from de Vries et al. (2006) has been used widely to measure the steepness of  
295 a dominance hierarchy, even though this measure is known to be negatively correlated with the  
296 percentage of unknown relationships (Klass and Cords, 2011) and unknown relationships are  
297 common in empirical datasets (Shizuka & McDonald, 2012). Already in their initial publication,  
298 de Vries et al. warned about interpreting results from matrices with “observational zeroes” (i.e.,  
299 unknown relationships). We argue that the problem arises because the David's Score treats  
300 unknown relationships as if both partners have the same ability to win competitive encounters. To  
301 address this limitation, de Vries et al. propose for each unknown dyad to use other behavioural  
302 observations outside of competitive interactions (“circumstantial observations”, de Vries et al.  
303 2006) to see whether an egalitarian relationship is a good approximation. For example, two  
304 individuals actively avoiding each other show an unresolved dominance relationships, in which  
305 case the unknown relationships can effectively be considered egalitarian. This is also true when

306 two individuals tolerate each other by spending time in close proximity without physical  
307 aggression, which means that a dominance relationship between the two is missing. However,  
308 when one individual actively avoids the other, because a clear dominance relationship was already  
309 established in the past, the dyad should not be treated as egalitarian, but how the dyad should be  
310 treated instead (in the mathematical calculation of the index) is unclear. In addition to this problem,  
311 the use of circumstantial observations to define dominance relationships is problematic, as it  
312 requires additional data which are not always collected (or properly codified). To minimize the  
313 effect that unknown relationships have on the steepness-slope we propose to base it on the Average  
314 Dominance Index rather than the David's Score, because the Average Dominance Index excludes  
315 unknown dyads from the calculation, preventing researchers from relying on circumstantial  
316 observations and reducing the deviation of steepness values when the percentage of unknown  
317 relationships increases (Table 2 and Figure 1).

318 Our results confirm the negative relationship between the steepness-slope based on the David's  
319 Score and the percentage of unknown relationships in a group as shown by Klass and Cords (2011).  
320 We show that when the Average Dominance Index is used in the steepness-slope, this greatly  
321 reduces the effect of unknown relationships on the steepness measure. Although the steepness-  
322 slope based on the Average Dominance Index shows a significant, positive trend with an increasing  
323 percentage of unknown relationships, this trend is considerably smaller than that caused by the  
324 David's Score. Also for the combination of bias (the intensity of trend) and variation in values,  
325 tested with the root mean squared deviation of the two indices, the Average Dominance Index  
326 performs between 5.4 and 3.6 times better than the David's Score, even though the use of the  
327 Average Dominance Index results in a greater variation in steepness-slope values. The larger  
328 variability of the steepness-slope when using the Average Dominance Index when dyads are

329 excluded could arise from the effect of missing dyads being larger and more random than the  
330 egalitarian, homogeneous treatment of unknown relationships in the David's Score.

331 In the computational model DomWorld, we show a) the same trend of the steepness-slope with the  
332 percentage of unknown relationships as in empirical data, b) the similarity of the two indices, the  
333 David's Score and the Average Dominance Index, when all relationships are known, and c) that  
334 even when combining bias and variability the Average Dominance Index performs better than the  
335 David's Score. We also show that the David's Score's corrected for the number of interactions in  
336 a dyad ( $D_{ij}$ ) performs similarly to its un-corrected version ( $P_{ij}$ ) regarding the bias of the steepness-  
337 slope with an increasing percentage of unknown relationships.

338 The range of values of the steepness-slopes differ for the two dominance indices. When the David's  
339 Score is used, steepness-slopes never exceed the value of one, which is the theoretical maximum  
340 of a perfectly linear dominance hierarchy. The exclusion of any random dyads does not result in  
341 steeper values as each unknown relationship is treated as perfectly egalitarian and therefore, only  
342 decreases the general steepness of the hierarchy. When the Average Dominance Index is used,  
343 however, the steepness-slopes may reach values higher than one because the exclusion of unknown  
344 relationships may cause the hierarchical structure to become steeper (Figure 1).

345 In this work, for the calculation of the steepness-slope we use the David's Score with its correction  
346 for the number of interactions in a dyad ( $D_{ij}$ ), as originally proposed by de Vries et al. (2006). We  
347 show that when the percentage of unknown relationships increases, the trend of the steepness-  
348 slope is virtually identical when we use the David's Score with the correction for the number of  
349 interactions in each dyad ( $D_{ij}$ ) or without it ( $P_{ij}$ ). We find that our use of  $D_{ij}$  returns a steepness-  
350 slope smaller than that of  $P_{ij}$ , similarly to what was found by de Vries et al. (2006). It is worth  
351 noting that, although we do not find a difference between our use of  $D_{ij}$  and  $P_{ij}$  in relation to



352 increasing percentages of unknown relationships, this might be caused by our treatment of  
353 unknown relationships. Specifically, we iteratively excluded interactions between selected dyads  
354 of individuals to increase the percentage of unknown relationships in our data, but the number of  
355 interactions in the dyads that are not excluded is not affected. Thus, the correction for the number  
356 of interactions in a dyad ( $D_{ij}$ ) does not have any effect. In some species, if groups with a higher  
357 percentage of unknown relationships also have a lower number of interactions in each dyad, this  
358 may still artificially reduce steepness-slope values when using  $D_{ij}$  instead of  $P_{ij}$ , as supposed by  
359 Balasubramaniam et al. (2013). However, we show that the percentage of unknown relationships  
360 alone cannot account for any difference between David's Score's  $D_{ij}$  and  $P_{ij}$  in the calculation of  
361 the steepness-slope.

362 Strauss and Holekamp (2019) proposed another modification of the David's Score, that considers  
363 previous interactions to the period analysed to better calculate David's Scores for each individual,  
364 and this slightly improves the performance of the index when unknown relationships increase. We  
365 do not test this modification in our present study, because we only test the methodology originally  
366 proposed. Although this modification is unlikely to significantly change the balance between the  
367 use of the Average Dominance Index and the David's Score, future work might consider  
368 investigating its effect on the performance of the David's Score in the steepness-slope.

369 In our present study, we reported results for two empirically relevant proportions of unknown  
370 relationships (25% as an average value, and 66.7% as the upper threshold: Shizuka & McDonald,  
371 2012). This choice was made to highlight the trend occurring in the steepness-slope in relation to  
372 unknown relationships. It was not meant to indicate specific steepness values at any particular  
373 percentage of unknown relationships. When the steepness-slope was calculated with the Average  
374 Dominance Index, the difference in values between the steepness-slope of the original group and

375 that of the group with an increased percentage of unknown relationship was not significantly  
376 different from zero until, on average for the four wild groups and the example group of DomWorld,  
377 30.2% unknown relationships. Even so, values varied widely among groups and over time, with  
378 Kubu in 2017 having the lowest proportion (20%) and Noha in 2016 the highest (43%). When the  
379 steepness-slope was calculated with the David's Score, differences in steepness values were  
380 significantly different from zero for all percentages of unknown relationships and all groups. Thus,  
381 specific indications are not possible because of the difference among groups in trend for steepness-  
382 slopes in relation to unknown relationships.

383 In future, other methodologies of measuring the steepness of hierarchy (such as proposed by  
384 Sánchez-Tójar et al. 2017), should be tested for their performance in relation to unknown  
385 relationships using a similar study design.

## 386 **Conclusion**

387 We show that the steepness-slope suffers less from unknown relationships when calculated based  
388 on the normalized Average Dominance Index than the normalized David's Score. Specifically, the  
389 performance of the two indices is almost identical when unknown relationships are absent, but  
390 with increasing percentages of unknown relationships the hierarchical steepness is substantially  
391 less biased and its overall standard error is substantially smaller when it is calculated with the  
392 normalized Average Dominance Index than with the normalized David's Score.

## 393 **Author Contributions**

394 **Tommaso Saccà:** Conceptualization, Software, Formal analysis, Writing - Original Draft.

395 **Gerrit Gort:** Supervision, Writing - Review & Editing. **Erica van de Waal:** Resources, Writing

396 - Review & Editing. **Charlotte K Hemelrijk:** Supervision, Writing - Review & Editing.

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437

438 **Table 1**

439 Two fictional competitive interaction matrices for four individuals.

**Interaction matrix I    Interaction matrix II**

	<b>A</b>	<b>B</b>	<b>C</b>	<b>D</b>		<b>A</b>	<b>B</b>	<b>C</b>	<b>D</b>
<b>A</b>	-	2	2	2	<b>A</b>	-	0	4	4
<b>B</b>	2	-	2	2	<b>B</b>	0	-	0	0
<b>C</b>	2	2	-	5	<b>C</b>	2	2	-	5
<b>D</b>	2	2	4	-	<b>D</b>	2	2	4	-

440 Two fictional competitive interaction matrices for four individuals (A, B, C and D) with winners

441 in rows and losers in columns. Left: egalitarian case of A and B (same number of wins and losses

442 in all dyads they are involved in); right: non-egalitarian case of A and B (unknown relationships

443 between A and B, A wins all interactions, B loses all interactions.)

Group (size)	Original # of unknown relationships (total, %)	Original steepness	normalized David's Score		Kendall's Rank Correlation	
			Median steepness (% change) at 25% unkn. relationships	Median steepness (% change) at 66.7% unkn. relationships	Average Tau ( $\pm$ SD)	Wilcox on signed rank test P-value
<b>KB2017 (6)</b>	1 (15, 7)	0.68	0.46 (-31.8) <sup>a</sup>	0.15 (-77.8)	-0.999 $\pm$ 0.008	< 0.001
<b>AK2011 (7)</b>	3 (21, 14)	0.55	0.45 (-17.5) <sup>b</sup>	0.13 (-76.0)	-0.998 $\pm$ 0.008	< 0.001
<b>NH2016 (8)</b>	3 (28, 11)	0.54	0.41 (-24.7)	0.12 (-77.5) <sup>d</sup>	-0.997 $\pm$ 0.004	< 0.001
<b>NH2011 (10)</b>	3 (45, 7)	0.75	0.52 (-30.9) <sup>c</sup>	0.14 (-80.6)	-0.999 $\pm$ 0.002	< 0.001
<b>DomWorld example (10)</b>	0 (45, 0)	0.74	0.45 (-39.2) <sup>c</sup>	0.12 (-83.9)	-1 $\pm$ 0.0004	< 0.001

normalized Average Dominance

Index

<b>KB2017 (6)</b>	1 (15, 7)	0.97	0.99 (+1.96) <sup>a</sup>	1.19 (+22.6)	0.43 ± 0.34	< 0.001
<b>AN2011 (7)</b>	3 (21, 14)	0.87	0.87 (-0.09) <sup>b</sup>	1.09 (+25.1)	0.43 ± 0.38	< 0.001
<b>NH2016 (8)</b>	3 (28, 11)	0.85	0.85 (-0.54)	0.99 (+16.6) <sup>d</sup>	0.31 ± 0.38	< 0.001
<b>NH2011 (10)</b>	3 (45, 7)	0.93	0.93 (+0.04) <sup>c</sup>	1.09 (+17.2)	0.35 ± 0.38	< 0.001
<b>DomWorld example (10)</b>	0 (45, 0)	0.76	0.77 (+0.95) <sup>c</sup>	0.86 (+12.9)	0.32 ± 0.40	< 0.001

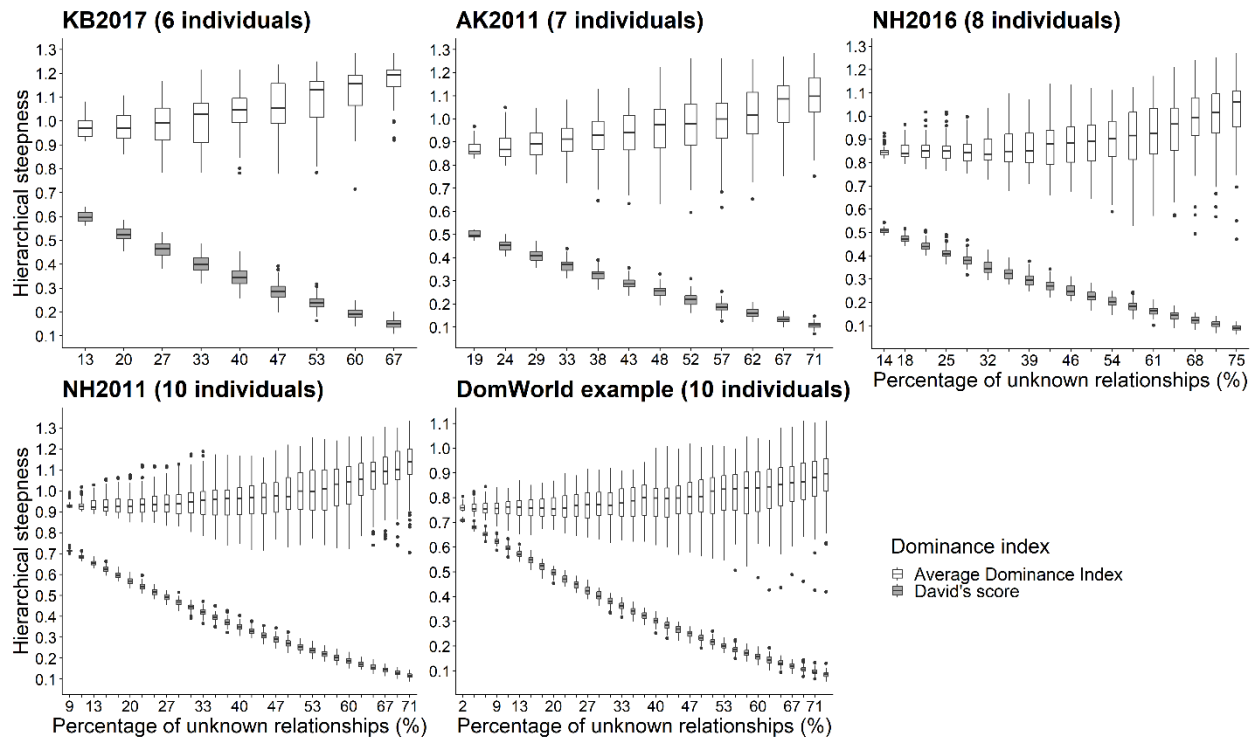
<sup>a</sup>Median steepness (% change) at 26.7% unknown relationships

<sup>b</sup>Median steepness (% change) at 23.8% unknown relationships

<sup>c</sup>Median steepness (% change) at 24.4% unknown relationships

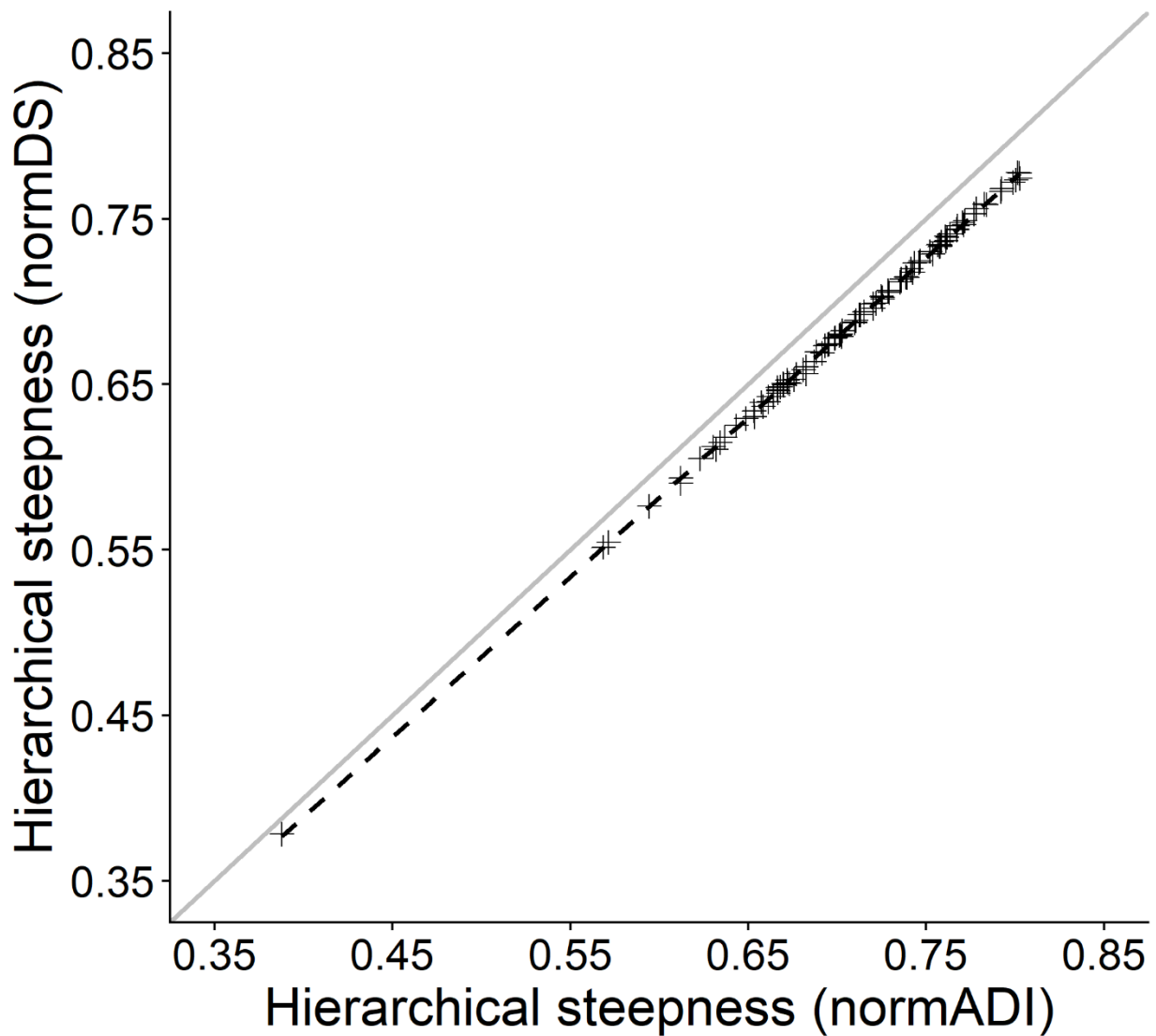
<sup>d</sup>Median steepness (% change) at 67.9% unknown relationships

446 Results for the Monte Carlo simulations for one-fourth and two-third unknown relationships in the  
447 social interaction matrix and Kendall Rank statistics for the correlation between the steepness-  
448 slope and the percentage of unknown relationships.



449 **Figure 1:** The steepness-slope (calculated with the Average Dominance Index and David's Score)  
 450 and the percentage of unknown relationships in matrices of social interaction of four group-years  
 451 of vervet monkeys (KB2017, AK2011, NH2016, NH2011) and one DomWorld run. Names of  
 452 group-years of vervet monkeys indicate the abbreviation (two letters) of the name of the group and  
 453 the year of observation. Group size is indicated in brackets for each group-year in empirical data  
 454 of groups of vervets and for a DomWorld example run (see Table 2). For the boxplots, edges of  
 455 boxes represent first and third quartile, internal lines represent median values, external lines show  
 456 minimal and maximal values, while dots depict outliers. (2-column fitting image)





457 **Figure 2:** The steepness-slope calculated with the normalized Average Dominance Index  
 458 (normADI, on the x axis) versus with the normalized David's Score (normDS, on the y axis) for  
 459 100 runs of DomWorld. The line in grey shows a regression line in which the use of the two indices  
 460 results in identical values for hierarchical steepness. (single column fitting image)