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Robust, Universal Tree Balance Indices

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Abstract.—Balance indices that quantify the symmetry of branching events and the compactness of trees are widely used to compare evolutionary processes or tree-generating algorithms. Yet, existing indices are not defined for all rooted trees, are unreliable for comparing trees with different numbers of leaves, and are sensitive to the presence or absence of rare types. The contributions of this article are twofold. First, we define a new class of robust, universal tree balance indices. These indices take a form similar to Colless' index but can account for population sizes, are defined for trees with any degree distribution, and enable meaningful comparison of trees with different numbers of leaves. Second, we show that for bifurcating and all other full m -ary cladograms (in which every internal node has the same out-degree), one such Colless-like index is equivalent to the normalized reciprocal of Sackin's index. Hence, we both unify and generalize the two most popular existing tree balance indices. Our indices are intrinsically normalized and can be computed in linear time. We conclude that these more widely applicable indices have the potential to supersede those in current use. [Cancer; clone tree; Colless index; Sackin index; species tree; tree balance.]

Tree balance indices—most notably those credited to Sackin (1972) and Colless (1982)—are widely used to describe speciation processes, compare cladograms, and assert the correctness of tree reconstruction methods (Shao and Sokal 1990; Mooers and Heard 1997; Fischer et al. 2021). Existing tree balance indices have several important flaws. First, they cannot be applied to any tree in which any node has only one descendant. Second, existing indices are unreliable for comparing trees with different numbers of leaves. Third, because they do not account for population sizes, these indices are sensitive to the omission or inclusion of rare types. The latter issue is, for example, a problem in oncology (Chkhaidze et al. 2019; Scott et al. 2020), where methods for determining and classifying evolutionary modes have clinical value (Davis et al. 2017; Maley et al. 2017).

Here, we develop a new class of robust, universal tree balance indices. Our definitions not only extend the tree balance concept and open up new applications but also unify the two main approaches to quantifying balance as proposed by Sackin and Colless. We describe several general advantages of our indices compared to those in current use.

MATERIALS AND METHODS

Rooted Trees

We consider exclusively rooted trees in which all edges are oriented away from the root (which will be topmost in our figures). This orientation defines a natural order on the tree, from top to bottom: edges descend from the root to the other *internal nodes* and finally to the terminal nodes or *leaves*. The *out-degree* of a node i , written $d^+(i)$, is the number of direct descendants, ignoring any subtrees in which all nodes have zero size. Internal nodes have out-degree at least one, whereas leaves have out-degree

zero. If all internal nodes have out-degree 1, then the tree is called *linear*. If all internal nodes have out-degree $m > 1$ then the tree is a *full m -ary tree*, and if $m = 2$ then it is also called *bifurcating* (such as Fig. 1a,b).

Some other tree topologies have particular names. A *caterpillar tree* (Fig. 1a) is a bifurcating tree in which every internal node except one has exactly one leaf. A *fully symmetric tree* (Fig. 1b) is such that every internal node with the same depth has the same degree or, equivalently, for each internal node i all the subtrees rooted at i are identical. A *star tree* (Fig. 1c) is a tree whose leaves are all attached to the root, which is the only internal node.

Node Sizes, Tree Magnitudes, and Leafy Trees

Although our definitions can be applied in other contexts, we will assume that nodes correspond to biological taxa or clones, and on this basis, we assign non-negative *node sizes*. If we know (or care) only whether each type is extant or extinct—as is typical in taxonomy—then we assign size zero to every node representing an extinct type, and size one otherwise. If nodes represent clones with known population sizes—as is often the case in studies of cancer and microbial evolution—then each node size is equal to the population size of the corresponding clone. The *magnitude* of a tree or subtree is then defined as the sum of its node sizes (we use magnitude here because a tree's size is conventionally defined as its number of nodes). We define a *leafy tree* as a rooted tree in which all internal nodes have size zero.

Cladograms, Taxon Trees, and Clone Trees

Tree types can also be defined in terms of what they represent. Following Podani (2013), we distinguish between two representations used in systematic biology.

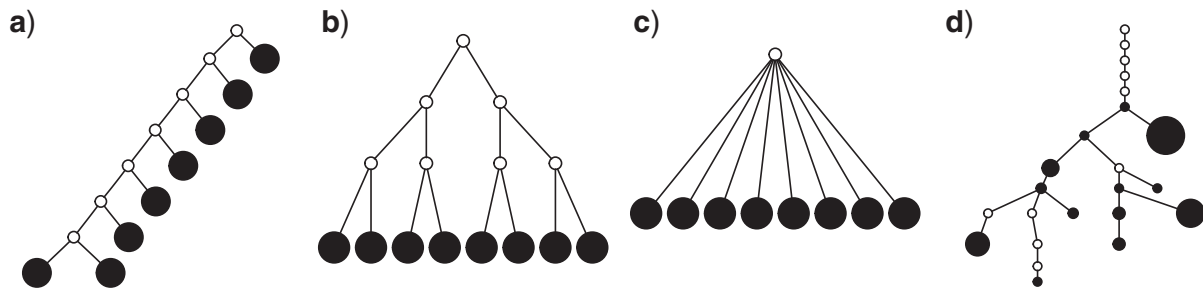


FIGURE 1. Contrasting trees. a) Caterpillar tree with $I_S=35$, $I_{S,norm}=1$, $I_C=21$, $I_{C,norm}=1$, $I_\Phi=56$, $I_{\Phi,norm}=1$. b) Fully symmetric bifurcating tree with $I_S=24$, $I_{S,norm}\approx 0.59$, $I_C=I_{C,norm}=0$, $I_\Phi=16$, $I_{\Phi,norm}\approx 0.29$. c) Star tree with $I_S=8$, $I_{S,norm}=0$, I_C and $I_{C,norm}$ undefined, $I_\Phi=I_{\Phi,norm}=0$. d) Clone tree of the lung tumor CRUK0065 in the TRACERx cohort (Jamal-Hanjani et al. 2017). In the clone tree, nodes represented by empty circles correspond to extinct clones, and the diameters of other nodes are proportional to the corresponding clone population sizes.

We define a *cladogram* as a rooted tree in which internal nodes represent hypothetical extinct ancestors, leaves represent extant biological taxa, and edges represent evolutionary relationships. This is equivalent to the synchronous cladogram definition of Podani (2013). Every cladogram is by definition a leafy tree, with a magnitude equal to its number of leaves. A common conception is that only bifurcating cladograms can be considered fully resolved. However, the linear two-node cladogram is appropriate for representing serial anagenesis (in which each descendant replaces its ancestor), while budding (in which an ancestor produces a descendant and remains extant) can give rise to cladogram nodes with an out-degree greater than two (Podani 2013). Hence, there is no restriction on cladogram node degrees. An extant ancestor is represented in a cladogram by a leaf stemming from the internal ancestor node, in which case, as Podani notes, “an ancestor is identical to an extant taxon connected directly to it.”

Alternatively, extant or known ancestors may be represented uniquely by internal nodes (like in a genealogy with overlapping generations). Such diagrams are known to organismal biologists as species trees or taxon trees, and to oncologists as clone trees. We define a *taxon tree* as a rooted tree in which all nodes represent biological taxa, and edges represent ancestor-descendant relationships. Similarly, a clone tree is defined as a rooted tree in which each node represents a clone (a set of cells that share alterations of interest due to common descent), and edges represent the chronology of alterations. Both taxon tree and clone tree fit the achronous tree definition of Podani (2013). Clone tree nodes can have any out-degree, including $d^+=1$, and each node—including internal nodes—can be associated with a non-negative size, as illustrated in Figure 1d.

When nodes are associated with sizes, the addition of subtrees comprising even vanishingly small nodes can change leaves into internal nodes and so substantially change the value of existing tree balance indices. This behavior is unsatisfactory because relatively small nodes typically represent either newly created types that have yet to experience evolutionary forces or types on the verge of extinction, and in either case convey negligible information about the mode of evolution. Data sets may

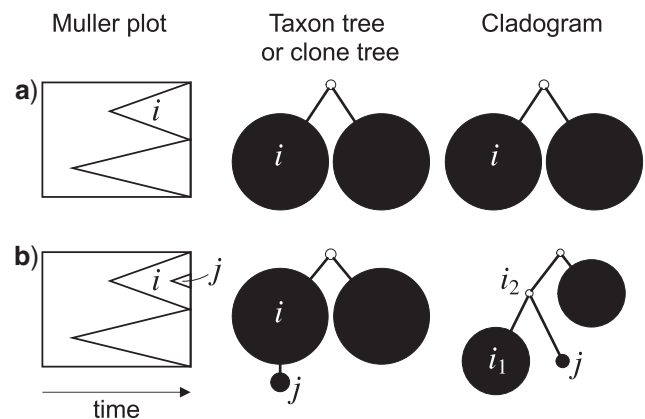


FIGURE 2. Muller plots (left column), taxon or clone trees (middle column), and cladograms (right column) representing evolution by splitting only (a) and both splitting and budding (b). In a Muller plot, polygons represent proportional subpopulation sizes (vertical axis) over time (horizontal axis), and each descendant is shown emerging from its parent polygon. In the trees, nodes represented by empty circles correspond to extinct types.

also omit rare types due to sampling error or because genetic sequencing methods have imperfect sensitivity (Turajlic et al. 2018).

The change due to the addition of terminal nodes is greater when the tree is a cladogram rather than a taxon or clone tree. For example, when a three-node, two-leaf tree (Fig. 2a) is augmented by adding a node j to a leaf i (Fig. 2b), the three original nodes retain their positions in the clone tree (middle column of Fig. 2), but in the cladogram (right column) node i becomes two nodes (i_1 and i_2), the larger of which is now further from the root (see Podani (2013) for further illustrations of this difference). As the size of the new node j is continuously reduced to zero, the clone tree changes continuously, whereas the cladogram undergoes an abrupt change of topology when the size of node j reaches zero. We conclude that the taxon tree or clone tree representation is more robust than the cladogram representation in the general case in which nodes are associated with sizes and ancestors can be extant. Also, an index that accounts for nonzero internal node sizes can be made more robust than one that does not. Accordingly, we will

define indices for the more general domain of clone trees and then obtain results for cladograms as a special case.

Existing Tree Balance Indices

The most widely used tree balance indices are in fact imbalance indices, such that more balanced trees are assigned smaller values. These indices were introduced to study cladograms; they take no account of node size, and, even after applying standard normalizations, they are appropriate only for comparing trees with equal numbers of leaves. The most popular are Sackin's index and Colless' index.

Sackin's index.—Let T be a tree with a set of leaves $L(T)$. For a leaf $l \in L(T)$, let v_l be the number of internal nodes between l and the root, which is included in the count. Then, the index credited to Sackin (1972) is

$$I_S(T) = \sum_{l \in L(T)} v_l.$$

For two bifurcating trees with the same number of leaves, a less balanced tree has higher values of v as the tree is in a sense less compact (compare trees a and b in Fig. 1).

Since the value tends to increase with the number of nodes, Shao and Sokal (1990) proposed normalizing I_S with respect to trees on $n > 2$ leaves by subtracting its minimum possible value for such trees and then dividing by the difference between the maximum and minimum possible values. The minimal I_S is reached on the star tree, such as tree c in Figure 1, and hence $\min_n(I_S) = n$. The maximum is attained on the caterpillar tree, such as tree a:

$$\max_n(I_S) = n - 1 + \sum_{v=1}^{n-1} v = n - 1 + n(n-1)/2 = (n-1)(n+2)/2.$$

The normalized index is then

$$I_{S,\text{norm}}(T) = \frac{I_S(T) - n}{(n+2)(n-1)/2 - n}.$$

This normalized index is not very satisfactory as a balance index because it fails to capture an intuitive notion of balance. For example, it is not obvious why a fully symmetric tree (b) should be considered less balanced than the star tree (c) in Figure 1, yet its $I_{S,\text{norm}}$ value is much larger. To address this issue, Shao and Sokal (1990) further suggested normalizing I_S relative to its extremal values among trees with the same number of internal nodes as well as the same number of leaves. But even then the index remains unreliable for comparing trees with different numbers of leaves. For example, the index is 1 for every caterpillar tree, yet long caterpillar trees are intuitively less balanced than short ones. The conventional I_S normalizations are not defined for trees containing linear parts. Moreover, since I_S does not account for node size, it is sensitive to the addition or removal of subtrees comprising relatively small nodes.

Colless' index.—For an internal node i of a bifurcating tree T , define n_{i_1} as the number of leaves of the left branch of the subtree rooted at i , and n_{i_2} as the number of leaves of the right branch. Then, the index defined by Colless (1982) is

$$I_C(T) = \sum_{i \in \tilde{V}(T)} |n_{i_1} - n_{i_2}|,$$

where $\tilde{V}(T)$ is the set of all internal nodes of T . The index can be normalized for the set of trees on $n > 2$ leaves by dividing by its maximal value, $\binom{n-1}{2}$, which is reached on the caterpillar tree (as in Fig. 1a).

Because Colless' index cannot be applied to multifurcating trees, Mir et al. (2018) recently introduced a family of Colless-like balance indices, including I_C as a special case. Each of these indices $C_{D,f}$ is determined by a weight function f , which assigns a size to each subtree as a function of its out-degree, and a dissimilarity function D . By definition of D , Colless-like indices are zero if and only if each internal node divides its descendants into subtrees of equal size. But since these indices are normalized by dividing by the maximal value for trees on the same number of leaves, they are unreliable for comparing trees with different numbers of leaves. In common with Sackin's index, the total cophenetic index I_Φ (Mir et al. 2013) (see Appendix), and other existing indices (surveyed by Fischer et al. (2021)), the Colless-like indices so far defined do not account for node sizes and can be applied only to trees in which all nodes have out-degree greater than one.

Desirable Properties of a Universal, Robust Tree Balance Index

Our aim is to derive a tree balance index J that is useful for classifying and comparing rooted trees that can have any distributions of node degrees and node sizes. Here, we specify four desirable properties that such an index should have. The first two axioms relate to extrema. We will call an index *universal* if it is defined for trees with any degree distribution and obeys these first two axioms. An index that conforms to the other three axioms—which are relevant only when nodes can have arbitrary sizes—will be called *robust*.

We will begin by introducing some additional notation (see also Table 1). For a tree T , we will use $V(T)$ to denote the set of all nodes of T , which we will abbreviate to V when the identity of the tree is unambiguous. Let $f(v) \geq 0$ denote the size of node v . Then, T_i denotes the subtree rooted at node i (i.e., the subtree that contains node i and all its descendants); S_i is the magnitude of T_i ; and S_i^* is the magnitude of T_i excluding its root:

$$S_i := \sum_{v \in V(T_i)} f(v); \quad S_i^* := \sum_{\substack{v \in V(T_i) \\ v \neq i}} f(v) = S_i - f(i).$$

We will use $\tilde{V}(T)$ or simply $\tilde{V}$ to denote the set of all internal nodes such that $\{i \in \tilde{V}\} := \{i \in V : S_i^* > 0\}$.

TABLE 1. Notation used throughout this article

Properties of a node i	
$d^+(i)$	Out-degree
$C(i)$	Set of children
$v(i)$	Depth
$f(i)$	Size
T_i	Subtree rooted at i
n_i	Number of leaves of T_i
S_i	Magnitude of T_i (sum of node sizes)
S_i^*	Magnitude of T_i excluding its root
g_i	Importance factor
p_{ij}	S_j/S_i^* , where $j \in C(i)$
W_i	Balance score
W_i^q	Balance score based on qH
h_i	Nonroot dominance factor
Sets of nodes	
V	All nodes
$\tilde{V}$	Internal nodes i such that $S_i^* > 0$
L	Leaves
Entropies and tree balance indices	
qH	Generalized entropy with parameter q
1H_b	Shannon entropy with base b
I_S	Sackin's index
I_C	Colless' index
I_Φ	Total cophenetic index
$C_{D,f}$	Colless-like index
$I_{S,gen}$	Generalized Sackin's index
$I_{C,gen}$	Generalized Colless' index
J^q	Tree balance index based on qH
J_S	Normalized inverse Sackin index
J^{1c}	A conservative tree balance index

Conventionally, a tree is considered maximally balanced only if every internal node splits its descendants into subtrees on the same number of leaves (Shao and Sokal 1990). We generalize this concept by requiring that every internal node splits its descendants into at least two subtrees of equal magnitude, as in Figure 3a. We call this the *equal splits* property, and we make it a necessary and sufficient condition for maximal balance.

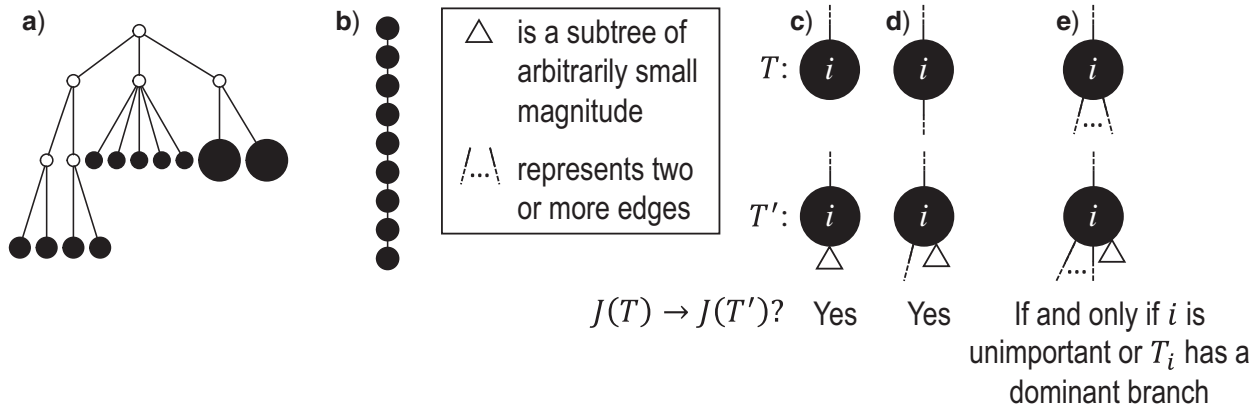


FIGURE 3. a) A tree in which each internal node has null size and splits its descendants into subtrees of equal magnitude, and hence $J = 1$. This tree can be considered balanced only according to an index that accounts for node size. b) A linear tree, for which $J = 0$. c–e) A robust, universal tree balance index J is insensitive to the addition of a subtree of arbitrarily small magnitude if it is added to a leaf (a) or a nonroot node with out-degree 1 (b), but not necessarily if the subtree is added to a nonroot node with greater out-degree (c).

Axiom 1 (Maximum value). $J(T) \leq 1$ for all trees T , and $J(T) = 1$ if and only if T has equal splits.

Another convention is that trees with relatively many internal nodes are considered highly imbalanced. According to this convention, linear trees (i.e., trees in which every node i has $d^+(i) \leq 1$, as in Fig. 3b) should be considered even less balanced than caterpillar trees. Also, given that balance implies branching, the most imbalanced split is one that assigns all descendants to one branch and none to any other branches. Hence our second desirable property:

Axiom 2 (Minimum value). $J(T) \geq 0$ for all trees T , and $J(T) = 0$ if and only if T is a linear tree.

Our third desirable property ensures that our index is insensitive to the properties of nodes that have relatively few descendants.

Axiom 3 (Insensitivity). Let T be a tree and l be one of its leaves. If we create a new tree T' from T by adding a subtree with finitely many nodes rooted at l then $J(T') \rightarrow J(T)$ as $S_l^*/\sum_{j \in \tilde{V}(T')} S_j^* \rightarrow 0$.

Our fourth axiom ensures that a linear section of a tree is regarded as a maximally unequal split.

Axiom 4 (Linear limit). Let T be a tree and $i \in \tilde{V}(T)$ with $d^+(i) = 1$. Let i_1 be the unique child of i . If we create a new tree T' from T by adding additional subtrees with finitely many nodes rooted at i then $J(T') \rightarrow J(T)$ as $S_{i_1}/S_i^* \rightarrow 1$.

Lastly, we require continuity with respect to varying node size:

Axiom 5 (Continuity). Suppose we create a new tree T' by selecting a node of tree T and changing the node's size from x to x' . Then $J(T') \rightarrow J(T)$ as $x' \rightarrow x$.

Alternative axioms are considered in the Appendix.

Sensitivity to Changes in Out-degree of Nonroot Nodes

By design, our definition of a robust tree balance index does not require insensitivity to the addition or removal of rare types in all cases. To see why, suppose we transform a tree T into T' by adding one or more subtrees of arbitrarily small magnitude, attached to a nonroot node $i \in V(T)$. As illustrated in Figure 3c–e, there are three topologically distinct cases to consider. If i is a leaf of T (Fig. 3c) or $d^+(i) = 1$ in T (Fig. 3d) then $J(T') \rightarrow J(T)$ due to Axioms 3 or 4. In the first case, i is an *unimportant* node, which we define to mean that $S_i^* / \sum_{j \in \tilde{V}} S_j^* \rightarrow 0$. In the second case, if i is not an unimportant node in T then T_i must have a *dominant branch*, meaning that i has a child i_1 such that $S_{i_1} / S_i^* \rightarrow 0$. The third case, when $d^+(i) \geq 2$ in T (Fig. 3e), is more complicated. If i is an unimportant node in T then $J(T') \rightarrow J(T)$ as $S_i^* / \sum_{j \in \tilde{V}} S_j^* \rightarrow 0$ in T' , by Axiom 3. If T_i in T has a dominant branch T_{i_1} in T then $J(T') \rightarrow J(T)$ as $S_{i_1} / S_i^* \rightarrow 1$ in T' , by Axiom 4. But if neither of those conditions hold then our axioms do not specify the size of the effect on J .

Although we could modify Axiom 4 so that J is always insensitive to the addition of relatively low-magnitude subtrees—thus increasing the index's robustness—we argue that this would undermine its utility as a tree balance index. The balance of a node can be conventionally defined as the extent to which it splits its descendants into multiple subtrees of equal magnitude. By this definition, the attachment of a new, relatively low-magnitude subtree to a perfectly balanced node will create an imbalance even as—in fact especially as—the magnitude of this new subtree, relative to the magnitude of the node's pre-existing descendants, approaches zero. Therefore, it is desirable for a tree balance index to be sensitive to certain changes in node degree, such that in the third scenario considered above, $J(T') \rightarrow J(T)$ if and only if i is an unimportant node or T_i has a dominant branch (Fig. 3e).

RESULTS

General Definition of Universal, Robust Tree Balance Indices

Our general definition depends on two continuous functions of subtree magnitudes:

- An *importance* factor $g: \mathbb{R}_{>0} \rightarrow \mathbb{R}_{>0}$ with $g(x) \rightarrow 0$ as $x \rightarrow 0$;
- A *balance score* W that assigns $W_i \in [0, 1]$ to each internal node i such that $W_i = 0$ if and only if $d^+(i) = 1$, and $W_i = 1$ if and only if i splits its descendants into at least two equal-magnitude subtrees.

To allow us to define W more rigorously, let $\mathcal{S}$ denote the set of vectors with positive components that sum to unity:

$$\mathcal{S} := \bigcup_{k \geq 1} \{(x_1, \dots, x_k) | x_1, \dots, x_k > 0, x_1 + \dots + x_k = 1\}.$$

Then, $W: \mathcal{S} \rightarrow [0, 1]$ is such that, for all $(x_1, \dots, x_k) \in \mathcal{S}$:

- (Associativity) For every permutation π , $W(x_1, \dots, x_k) = W(x_{\pi(1)}, \dots, x_{\pi(k)})$;
- (Maximum value) $W(x_1, \dots, x_k) = 1$ if and only if $k > 1$ and $x_1 = \dots = x_k$;
- (Minimum value) $W = 0$ if and only if $\max(x_1, \dots, x_k) = 1$;
- (Continuity) W is a continuous function with respect to each of its arguments.

We then define a balance index in terms of subtree magnitudes as

$$J := \frac{1}{\sum_{k \in \tilde{V}} g_k} \sum_{i \in \tilde{V}} g_i W_i, \quad (1)$$

where $W_i = W(S_{i_1} / S_i^*, \dots, S_{i_p} / S_i^*)$, $g_i = g(S_i^* / \sum_{j \in \tilde{V}} S_j^*)$, and $i_1, \dots, i_p$ are the children of node i (see Table 1 for a recap of notation). A short proof that this type of index satisfies our five axioms for robustness and universality (Axioms 1–5) is presented in the Appendix.

The balance score W in Equation 1 measures the extent to which an internal node splits its descendants into equal-magnitude subtrees. The importance factor g assigns more weight to nodes that are the roots of large subtrees. In biological terms, this means giving more weight to types that have more descendants. Sackin's and Colless' indices similarly assign more weight to nodes that have more descendant leaves or are closer to the root. Mooers and Heard (1997) have argued that it is reasonable to put more weight on nodes deeper within the tree because “those nodes are the most informative, as the subclades they define are older and therefore sample longer periods of evolutionary time.”

A Specific Index Based on the Shannon Entropy

In defining a specific index, we start by opting for the simplest importance factor function: $g(x) = x$. The role of the balance score function W is to quantify the extent to which a set of objects (specifically subtrees) have equal magnitude. A well-known index that satisfies the necessary conditions is the normalized Shannon entropy.

Assume a population is partitioned into $n \in \mathbb{N}$ types, with each type i accounting for a proportion p_i . Then, the Shannon entropy with base b is defined as ${}^1H_b := -\sum_{i=1}^n p_i \log_b p_i$. If all types have equal frequencies $p_i = 1/n$, then ${}^1H_b = \log_b n$. If the types have unequal sizes, then ${}^1H_b < \log_b n$. And if the abundance is mostly concentrated on one type j , such that $p_j \rightarrow 1$, then ${}^1H_b \rightarrow 0$.

Let $C(i)$ denote the set of children (immediate descendants) of a node i , and for $j \in C(i)$ let $p_{ij} := S_j / S_i^*$ denote the relative magnitude of subtree T_j compared to all subtrees attached to i .



There are of course many alternative options for W . For example, Colless' index can be generalized to define a robust, though not universal, tree balance index on the domain of bifurcating trees (see Appendix). Since the Shannon entropy belongs to families of generalized

entropies (Rényi 1961; Chao et al. 2014) parameterized by $q > 0$, the above reasoning can be generalized to define a balance score W^q , and hence a robust, universal balance index J^q , for every $q > 0$ (see Appendix). Other candidates for W include one minus the variance of the proportional subtree magnitudes or one minus the mean deviation from the median (Mir et al. 2018). We prefer W^1 mostly because, as we shall show, it is the only function for which Equation 4 is a generalization of the normalized inverse Sackin index.

Relationship with Colless' Index

Like Colless' index and Colless-like indices as previously defined, our new family of tree balance indices is based on the intuitive idea of assigning a value to each internal node, summing these values, and then normalizing the sum. A Colless-like index in the sense of Mir et al. (2018) depends on a function $f: \mathbb{N} \rightarrow \mathbb{R}_{\geq 0}$, which assigns node sizes, and a dissimilarity score $D: \mathcal{R} \rightarrow \mathbb{R}_{\geq 0}$, where $\mathcal{R}$ is the set of non-null real vectors. Before normalization, such an index has the form

$$C_{D,f} = \sum_{i \in \tilde{V}} D(\delta_f(T_{i_1}), \dots, \delta_f(T_{i_k})),$$

where $\{i_1, \dots, i_k\}$ are the children of node i . The function δ_f assigns a size to each subtree by summing the node sizes: $\delta_f(T) = \sum_{j \in V(T)} f(d^+(j))$. Neglecting the initial normalizing factor, our general definition (Equation 1) has a similar form and can be considered Colless-like in only a slightly broader sense. Our definition nevertheless differs in two important ways.

First, whereas the unbounded dissimilarity index D measures both node imbalance and importance and is undefined for nodes with out-degree one, we split these two roles into a normalized balance score W and an unbounded importance factor g , and we assign a W value (specifically zero) to nodes with out-degree one. This difference enables us to extend the balance index definition to trees with any degree distribution. It also makes it easy to normalize our indices for any tree, simply by dividing by the sum of the important factors. Furthermore, our normalization is universal, rather than being based on comparison with other trees with the same number of leaves. For example, our J^q indices judge long caterpillar trees less balanced than short ones (Fig. 5a), whereas Sackin's index, Colless' index, and the total cophenetic index consider all caterpillar trees on more than two leaves equally imbalanced.

Second, instead of assigning a size to each node as a function of its out-degree, we associate a node's size with the size of the biological population it represents. This ensures that our indices can be made reliably robust by including population size data.

Relationship with Sackin's Index

The sum $\sum_{k \in \tilde{V}} n_k$ is just another way of expressing Sackin's index (summing over internal nodes instead

of leaves). Therefore, J in Equation 4 is essentially a weighted Sackin index (with each term in the sum weighted by the balance score W) divided by the unweighted Sackin index. In the special, important case of full m -ary leafy trees (including full m -ary cladograms), the weighted sum in J^1 (Equation 5) simplifies yet further. Let $\mathcal{T}_{n,m}^*$ denote the set of all trees on n leaves such that all internal nodes have the same out-degree $m > 1$, every internal node has null size, and all leaf sizes are equal. Then, we obtain a remarkably simple relationship between J^1 and Sackin's index:

Proposition 6. Let T be a tree on n leaves with $d^+(i) = m > 1$ and $f(i) = 0$ for every internal node i . Then

$$J^1(T) = \frac{{}^1H_m(T)S(T)}{I_{S,\text{gen}}(T)},$$

where ${}^1H_m(T)$ is the Shannon entropy (base m) of the proportional node sizes, $S(T)$ is the magnitude of T , and $I_{S,\text{gen}}(T) := \sum_{i \in \tilde{V}(T)} S_i^*$. If additionally all leaves of T have the same size (so $T \in \mathcal{T}_{n,m}^*$) then

$$J^1(T) = \frac{\min_{n,m} I_S}{I_S(T)} = \frac{n \log_m n}{I_S(T)}, \quad (6)$$

where $\min_{n,m} I_S$ is the minimum I_S value of trees in $\mathcal{T}_{n,m}^*$.

The above result is somewhat surprising as it unifies our Colless-like index, which can be viewed as a weighted average of internal node balance scores, and Sackin's index, which is the sum of all leaf depths. A short proof of Proposition 6 is presented in the Appendix. The converse result, which is also proved in the Appendix, justifies our choice of W^1 instead of alternative balance score functions:

Proposition 7. Let J be a tree balance index such that

$$J(T) = \frac{1}{\sum_{k \in \tilde{V}} n_k} \sum_{i \in \tilde{V}} n_i W\left(\frac{n_{i_1}}{n_i}, \dots, \frac{n_{i_{p(i)}}}{n_i}\right),$$

where $i_1, \dots, i_{p(i)}$ are the children of node i , and W is a balance score satisfying the conditions stated before Equation 1. Suppose that for all trees $T \in \mathcal{T}_{n,m}^*$, $J(T) = n \log_m n / I_S(T)$. Then, $W = W^1$.

The right-hand side of Equation 6 incidentally provides an alternative way of normalizing Sackin's index on full m -ary leafy trees, including the bifurcating cladograms on which the index was originally defined. This normalized inverse Sackin index, which we can define as $J_S := n \log_m n / I_S$, provides a more satisfactory way of comparing trees that differ in their node degrees or leaf counts. $J_S = 1$ if and only if the tree has minimal depth given m , which is equivalent to being fully symmetric, and so J_S is a sound tree balance index in the sense defined by Mir et al. (2018) (see Appendix for a proof). For $m > 1$, we have $J_S > 0$ but $\min J_S \rightarrow 0$ as $n \rightarrow \infty$, which makes sense because trees with more leaves can be

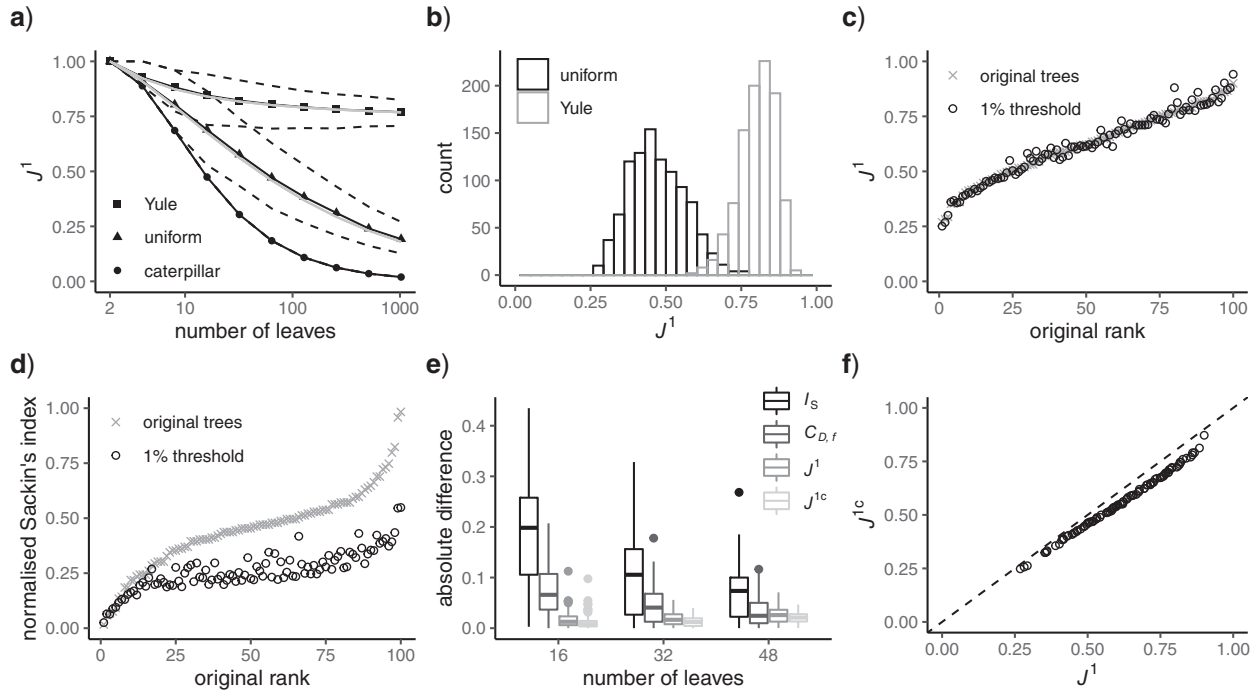


FIGURE 5. a) J^1 values for caterpillar trees and random trees generated from the Yule and uniform models (1000 trees per data point). All internal nodes have null size and all leaves have equal size. Solid black curves are the means; dashed curves are the 5th and 95th percentiles; and gray curves are $n \log_2 n$ divided by the corresponding expectation of I_S (where n is the number of leaves). b) J^1 distributions for random trees on 64 leaves generated from the Yule and uniform models (1000 trees per model). c) J^1 values for 100 random trees on 16 leaves, before and after applying a 1% sensitivity threshold. These random trees were generated from the alpha-gamma model with $\alpha \sim \text{Unif}(0, 1)$ and $\gamma \sim \text{Unif}(0, \alpha)$. d) $I_{S, \text{norm}}$ values for the same set of random trees. e) Absolute change in normalized index values due to applying a 1% sensitivity threshold. Results are based on 100 random trees for each number of leaves, generated as in (c) and (d). $C_{D,f}$ here is the Colless-like index with $f(n) = \ln(n+e)$ and D is the mean deviation from the median, as recommended by Mir et al. (2018). f) Values of J^{1c} versus J^1 for random multifurcating trees on 16 leaves, with node sizes drawn from a continuous uniform distribution. The dashed reference line has slope 1.

made less balanced. In particular, when T is a caterpillar tree on $n \geq 2$ leaves,

$$J_S(T) = \frac{2n \log_2 n}{(n-1)(n+2)},$$

as illustrated in Figure 5a. The definition of J_S can be naturally extended to the case $m \leq 1$ by setting $J_S(T) := 0$ if T is linear or has only one node. From this point of view, J^1 (a Colless-like index) is a generalization of J_S (the normalized reciprocal of Sackin's index) to the domain of trees with arbitrary degree distributions and arbitrary node sizes.

Distributions under the Yule and Uniform Models

An immediate corollary of Proposition 6 is that J^1 can be used to test whether a set of full m -ary cladograms is consistent with a particular tree-generating model, with exactly the same sensitivity as Sackin's index. For example, Figure 5a,b shows J^1 distributions for random bifurcating trees in $\mathcal{T}_{n,2}^*$ generated from the Yule and uniform models. These two distributions have insignificant overlap when the trees have at least a few dozen leaves.

Kirkpatrick and Slatkin (1993) showed that the expectation of I_S for the Yule model is

$$\mathbb{E}_{\text{Yule}}(I_S) = 2n \sum_{i=2}^n \frac{1}{i} = 2n \ln n + (2\gamma - 2)n + o(n),$$

where γ is Euler's constant and n is the number of leaves. Mir et al. (2013) have shown that the expectation of I_S for the uniform model is

$$\mathbb{E}_{\text{Unif}}(I_S) = n \left(\frac{(2n-2)!!}{(2n-3)!!} - 1 \right) = n \left(\frac{(2n-2)(2n-4)\dots(4)(2)}{(2n-3)(2n-5)\dots(3)(1)} - 1 \right),$$

which approaches $\sqrt{\pi} n^{3/2}$ as the number of leaves n approaches infinity (Blum et al. 2006; King and Rosenberg 2021). Consistent with Proposition 6, we find that for random trees in $\mathcal{T}_{n,2}^*$ generated by either the Yule or the uniform model, a good approximation to the J^1 mean is $n \log_2 n$ divided by the corresponding expectation of I_S (gray curves in Fig. 5a). As $n \rightarrow \infty$, these approximations approach $1/(2 \ln 2) \approx 0.72$ and zero for the Yule and uniform models, respectively.

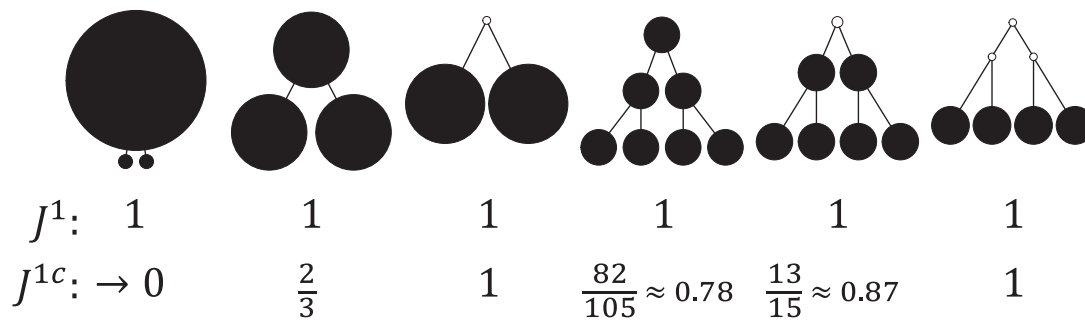


FIGURE 6. Example values of J^1 versus the conservative tree balance index J^{1c} . The latter index takes account of the size of each internal node, relative to the sum of its descendant node sizes.

Robustness when Applied to Random Trees

To test the robustness of J^1 , we generated random multifurcating trees with node sizes drawn from a continuous uniform distribution and then compared J^1 values for these trees before and after applying a 1% sensitivity threshold. In the latter case, whenever the combined frequency of a clone and its descendants was below 1%, we merged the corresponding subtree with the clone's parent, to simulate imperfect detection of rare types. As expected, the J^1 values for the two sets of trees were highly similar, with a median absolute difference of only 0.01 for trees that initially had 16 leaves (Fig. 5c). In contrast, the median absolute difference in the normalized Sackin's index for the same two sets of trees (after resolving any linear parts in the manner of Fig. 2) was 0.20 (Fig. 5d), confirming that J^1 is much more robust to the omission of rare types.

As the number of leaves per tree increases, indices such as Sackin's index and the Colless-like index recommended by Mir et al. (2018) become more robust to the removal of rare types (Fig. 5e). Like J^1 , these previously defined indices give more weight to nodes nearer the root. In larger trees, the nodes near the root tend to have large numbers of descendant leaves. It follows that removing a random sample of nodes from near the tips of the tree is likely to have only a modest effect on balance, as the tree's core structure is preserved. In our results, this effect outweighs an increase in the proportion of nodes removed (a median of 7%, 19%, and 24% of nodes were removed from trees that originally had 16, 32, and 48 leaves, respectively, by applying the 1% sensitivity threshold). Therefore the robustness benefit of J^1 is more pronounced in trees with fewer leaves.

Comparison with a Conservative Tree Balance Index

We additionally investigated the robustness of an alternative new tree balance index J^{1c} , defined as

$$J^{1c} := \frac{1}{\sum_{k \in \tilde{V}} S_k^*} \sum_{i \in \tilde{V}} S_i^* \frac{S_i^*}{S_i} W_i^1.$$

J^{1c} —which we denoted J^1 in a previous paper (Noble et al. 2022)—conforms to an alternative set of axioms that

define what we call a *conservative* tree balance index. This index is maximal not for all trees with equal splits, but only for leafy trees with equal splits (see Appendix for details).

An advantage of J^{1c} is that, unlike J^1 , it is always insensitive to adding relatively low-magnitude subtrees to the root of the tree. Nevertheless, as the number of nodes increases, the difference between J^1 and J^{1c} rapidly diminishes, unless the root node is disproportionately large (Fig. 6). For example, when J^1 and J^{1c} are applied to random multifurcating trees on 16 leaves, with node sizes drawn from a continuous uniform distribution, the linear correlation between the two indices is 0.998 (J^{1c} is approximately 10% smaller than J^1 in this case; Fig. 5f). Accordingly, we find that J^{1c} is only slightly more robust than J^1 to the removal of rare types when applied to reasonably large random trees (Fig. 5e). For most practical purposes, we see no strong reason to favor J^{1c} over the simpler index J^1 .

Resolution Power

Mir et al. (2013) have argued that a useful tree balance index should have good resolution power, meaning a low probability of assigning the same value to two trees with the same number of leaves, chosen uniformly at random. Proposition 6 implies that, when applied to full m -ary leafy trees with equally sized leaves, J^1 has the same resolution power as Sackin's index.

Correlations with Pre-existing Indices

To compare J^1 to Sackin's index, a Colless-like index, and the total cophenetic index (defined in the Appendix) on a diverse set of trees, we generated 2000 random multifurcating leafy trees on 100 equally sized leaves using the alpha-gamma model (Chen et al. 2009) via the R package *CollessLike* (Mir et al. 2018). As shown in Figure 7, our new balance index correlates negatively with the previously defined imbalance indices on this set of random trees, indicating that it captures a similar notion of balance. The strongest correlation is between J^1 and the total cophenetic index (Spearman's $\rho = -0.84$ for all trees, and $\rho = -0.97$ for trees with a mean out-degree

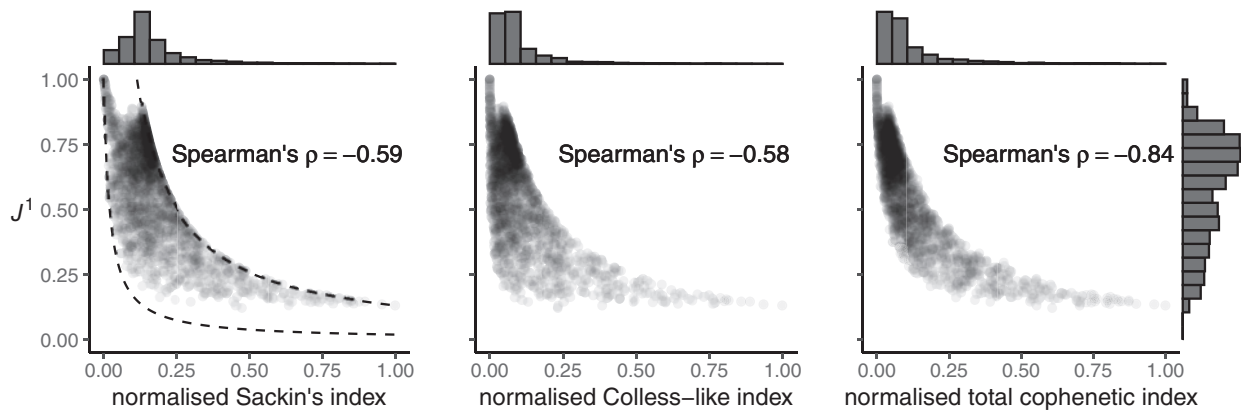


FIGURE 7. Scatter plots of J^1 versus normalized Sackin's, Colless-like, and total cophenetic indices for 2000 random multifurcating leafy trees with 100 equally sized leaves. Histograms in the margins show the marginal distributions. Dashed reference curves in the first panel are obtained by substituting $I_{S,\text{norm}}$ into Equation 6 with $n=100$ and $m=2$ (upper curve) or $m=100$ (lower curve). We use the Colless-like index with $f(n)=\ln(n+e)$ and D the mean deviation from the median, as recommended by Mir et al. (2018). Normalization of each index other than J^1 depends only on the number of leaves and so does not affect correlations. Trees were generated from the alpha-gamma model with $\alpha \sim \text{Unif}(0, 1)$ and $\gamma \sim \text{Unif}(0, \alpha)$.

greater than 3). The marginal histograms in Figure 7 additionally show that more than 85% of these random trees have balance values less than 0.25 according to the previously defined indices, whereas J^1 values are more evenly distributed between zero and one, with mean and median approximately equal to 0.6.

Sensitivity to Certain Changes in Node Degree

As explained in the Methods section, we consider it desirable for tree balance indices to be sensitive to certain changes in node degree. In J^1 this sensitivity arises because, in the calculation of the node balance score, the node out-degree features as the base of the logarithm. For example, consider a star tree T with $l > 1$ leaves each of size $f_0 > 0$. Suppose we add to the root another $n-l$ leaves, each of size $x > 0$. If $x=f_0$ then $J^1(T)=1$ since all the leaves have the same size. Otherwise

$$J^1(T) = - \left[l \frac{f_0}{lf_0 + (n-l)x} \log_n \left(\frac{f_0}{lf_0 + (n-l)x} \right) + (n-l) \frac{x}{lf_0 + (n-l)x} \log_n \left(\frac{x}{lf_0 + (n-l)x} \right) \right].$$

As x decreases from f_0 towards zero, $J^1(T)$ decreases monotonically to account for the growing loss of balance. And as $x \rightarrow 0$, so $J^1(T) \rightarrow \log_n l$. If we then remove these vanishingly small leaves, the value of $J^1(T)$ will jump from $\log_n l$ back to 1 because the remaining leaves are of equal size. The sensitivity of J^1 to such changes in node degree is thus a straightforward consequence of the conventional notion of node balance. The size of the jump in J^1 is at most $1 - \log_3 2 \approx 0.37$, and it approaches zero as $l/n \rightarrow 1$ (i.e., when the new nodes are relatively few). The analyses shown in Figure 5e,f show

that such discontinuities do not compromise the overall robustness of J^1 to the removal of rare types.

Implementation and Algorithmic Complexity

Assuming the identity of the root is known, our new indices can be computed from an adjacency matrix in $\mathcal{O}(N)$ time, where N is the number of nodes (or the number of edges plus one). Subtree magnitudes are computed via depth-first search, which takes linear time, and the computation of the balance index takes at most $\sum_{i=1}^N |\text{Adj}(i)| = N-1$ steps, where $\text{Adj}(i)$ is the adjacency list of node i . Efficient R code for calculating J^q is shared in an online repository (Noble and Lemant 2021).

DISCUSSION

Here, we have defined a new class of tree balance index that unifies, generalizes, and in various ways improves upon previous definitions. Even when restricted to the tree types on which pre-existing indices are defined, our indices enable a more meaningful comparison of trees with different degree distributions or different numbers of leaves. Due to these advantages, our indices have the potential to supersede those in current use.

Our indices also enable important new applications. A challenge in comparing simulated phylogenies and trees inferred from data is that the former are exact, whereas the latter are often incomplete (Scott et al. 2020). In oncology, for example, it has been shown that whether or not a rare tumor clone is detected depends on both methodology and chance (Turajlic et al. 2018). Our balance indices largely solve this problem as they are insensitive to the omission of rare types, as demonstrated briefly here and more comprehensively in a companion paper (Noble et al. 2022).

Because of its unique relationship with Sackin's index, we especially recommend J^1 —a weighted average of the normalized entropies of the internal nodes—as defined in general by Equation 3 and more simply for cladograms by Equation 5. Given that Sackin's index has been well studied, it is convenient that J^1 inherits some of the properties of that index when applied to full m -ary cladograms, including its relatively high sensitivity in distinguishing between alternative tree-generating models (Kirkpatrick and Slatkin 1993; Agapow and Purvis 2002). Within our framework, Sackin's index is seen not as a general balance index but rather as a normalizing factor, which works as a balance index only in the special case of full m -ary leafy trees (for which the numerator of J^1 is independent of tree topology).

Proposition 6 implies that determining the precise moments of J^1 for a model that generates full m -ary leafy trees is equivalent to determining the moments of the reciprocal of Sackin's index. Figure 7 suggests that J^1 has interesting relationships with other indices such as the total cophenetic index. These are promising areas for further investigation.

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REFERENCES

- Agapow P.M., Purvis A. 2002. Power of eight tree shape statistics to detect nonrandom diversification: a comparison by simulation of two models of cladogenesis. *Syst. Biol.* 51(6): 866–872.
- Blum M.G.B., François O., Janson S. 2006. The mean, variance and limiting distribution of two statistics sensitive to phylogenetic tree balance. *Ann. Appl. Prob.* 16(4): 2195–2214.
- Chao A., Chiu C.-H., Jost L. 2014. Unifying species diversity, phylogenetic diversity, functional diversity, and related similarity and differentiation measures through hill numbers. *Annu. Rev. Ecol. Syst.* 45(1):297–324.
- Chen B., Ford D., Winkler M. 2009. A new family of Markov branching trees: the alpha-gamma model. *Electron. J. Probab.* 14:400–430.
- Chkhaidze K., Heide T., Werner B., Williams M.J., Huang W., Caravagna G., Graham T.A., Sottoriva A. 2019. Spatially constrained tumour growth affects the patterns of clonal selection and neutral drift in cancer genomic data. *PLoS Comput. Biol.* 15(7):e1007243.
- Colless D.H. 1982. Review of phylogenetics: the theory and practice of phylogenetic systematics. *Syst. Zool.* 31(1):100–104.
- Davis A., Gao R., Navin N. 2017. Tumor evolution: linear, branching, neutral or punctuated? *Biochim. Biophys. Acta* 1867(2): 151–161.
- Fischer M., Herbst L., Kersting S., Kühn L., Wicke K. 2021. Tree balance indices: a comprehensive survey. *arXiv preprint arXiv:2109.12281*.
- Jamal-Hanjani M., Wilson G.A., McGranahan N., Birkbak N.J., Watkins T.B.K., Veeriah S., Shafi S., Johnson D.H., Mitter R., Rosenthal R., Salm M., Horswell S., Escudero M., Matthews N., Rowan A., Chambers T., Moore D.A., Turajlic S., Xu H., Lee S.-M., Forster M.D., Ahmad T., Hiley C.T., Abbosh C., Falzon M., Borg E., Marafioti T., Lawrence D., Hayward M., Kolvekar S., Panagiotopoulos N., Janes S.M., Thakrar R., Ahmed A., Blackhall F., Summers Y., Shah R., Joseph L., Quinn A.M., Crosbie P.A., Naidu B., Middleton G., Langman G., Trotter S., Nicolson M., Remmen H., Kerr K., Chetty M., Gomersall L., Fennell D.A., Nakas A., Rathinam S., Anand G., Khan S., Russell P., Ezhil V., Ismail B., Irvin-Sellers M., Prakash V., Lester J.F., Kornaszewska M., Attanoos R., Adams H., Davies H., Dentre S., Taniere P., O'Sullivan B., Lowe H.L., Hartley J.A., Iles N., Bell H., Ngai Y., Shaw J.A., Herrero J., Szallasi Z., Schwarz R.F., Stewart A., Quezada S.A., Le Quesne J., Van Loo P., Dive C., Hackshaw A., Swanton C. 2017. Tracking the evolution of non-small-cell lung cancer. *N. Engl. J. Med.* 376(22):2109–2121.
- King M.C., Rosenberg N.A. 2021. A simple derivation of the mean of the Sackin index of tree balance under the uniform model on rooted binary labeled trees. *Math. Biosci.* 342:108688.
- Kirkpatrick M., Slatkin M. 1993. Searching for evolutionary patterns in the shape of a phylogenetic tree. *Evolution* 47(4):1171–1181.
- Maley C.C., Aktipis A., Graham T.A., Sottoriva A., Boddy A.M., Janiszewska M., Silva A.S., Gerlinger M., Yuan Y., Pienta K.J., Anderson K.S., Gatenby R., Swanton C., Posada D., Wu C.-I., Schiffman J. D., Shelley Hwang E., Polyak K., Anderson A.R.A., Brown J.S., Greaves M., Shibata D. 2017. Classifying the evolutionary and ecological features of neoplasms. *Nat. Rev. Cancer* 17(10): 605–619.
- Mir A., Rosselló F., Rotger L.A. 2013. A new balance index for phylogenetic trees. *Math. Biosci.* 241(1):125–136.
- Mir A., Rotger L., Rosselló F. 2018. Sound Colless-like balance indices for multifurcating trees. *PLoS One* 13(9):e0203401.
- Mooers A.O., Heard S.B. 1997. Inferring evolutionary process from phylogenetic tree shape. *Q. Rev. Biol.* 72(1):31–54.
- Noble R., Lemant J. 2021. RUTreeBalance: robust, universal tree balance indices, 2021. <https://zenodo.org/badge/latestdoi/399934945>.
- Noble R., Burri D., Le Sueur C., Lemant J., Viossat Y., Kather J.N., Beerenwinkel N. 2022. Spatial structure governs the mode of tumour evolution. *Nat. Ecol. Evol.* 6(2):207–217.
- Podani J. 2013. Tree thinking, time and topology: comments on the interpretation of tree diagrams in evolutionary/phylogenetic systematics. *Cladistics* 29(3):315–327.
- Rényi A. 1961. On measures of entropy and information. *Proceedings of the Fourth Berkeley Symposium on Mathematical Statistics and Probability, Volume 1: Contributions to the Theory of Statistics*, vol. 4; University of California Press. p. 547–562.
- Sackin M.J. 1972. "Good" and "bad" phenograms. *Syst. Biol.* 21(2):225–226.
- Scott J.G., Maini P.K., Anderson A.R.A.A., Fletcher A.G. 2020. Inferring tumor proliferative organization from phylogenetic tree measures in a computational model. *Syst. Biol.* 69(4):623–637.
- Shao K.-T., Sokal R.R. 1990. Tree balance. *Syst. Zool.* 39(3):266.
- Turajlic S., Xu H., Litchfield K., Rowan A., Horswell S., Chambers T., O'Brien T., Lopez J.I., Watkins T.B.K., Nicol D., Stares M., Challacombe B., Hazell S., Chandra A., Mitchell T.J., Au L., Eichler-Jonsson C., Jabbar F., Soultati A., Chowdhury S., Rudman S., Lynch J., Fernando A., Stamp G., Nye E., Stewart A., Xing W., Smith J.C., Escudero M., Huffman A., Matthews N., Elgar G., Phillimore B., Costa M., Begum S., Ward S., Salm M., Boeing S., Fisher R., Spain L., Navas C., Grönroos E., Hobor S., Sharma S., Aurangzeb I., Lall S., Polson A., Varia M., Horsfield C., Fotiadis N., Pickering L., Schwarz R.F., Silva B., Herrero J., Luscombe N.M., Jamal-Hanjani M., Rosenthal R., Birkbak N.J., Wilson G.A., Pipek O., Ribli D., Krzystanek M., Csabai I., Szallasi Z., Gore M., McGranahan N., Van Loo P., Campbell P., Larkin J., Swanton C. 2018. Deterministic evolutionary trajectories influence primary tumor growth: TRACERx renal. *Cell* 173(3):595–610.e11.

APPENDIX

Definition of the Total Cophenetic Index

The cophenetic value $\phi(k, l)$ of a pair of leaves (k, l) is the depth of their lowest common ancestor. The total cophenetic index (Mir et al. 2013) is then the sum of the cophenetic values over all pairs of leaves:

$$I_{\Phi}(T) = \sum_{N-n+1 \leq k < l \leq n} \phi(k, l),$$

where N is the number of nodes and n the number of leaves. As in Sackin's index, the principle is that an unbalanced tree stretches more than a balanced tree. Being explicitly defined for all multifurcating trees, the total cophenetic index permits meaningful comparison of any two multifurcating trees on the same number of leaves.

For trees on $n > 2$ leaves, the minimum of the total cophenetic index is reached on the star tree, with $\min_n(I_{\Phi}) = 0$. The maximum is attained on the caterpillar tree:

$$\begin{aligned} \max_n(I_{\Phi}) &= \sum_{k=2}^{n-1} \sum_{l=1}^{k-1} m = \sum_{k=2}^{n-1} \frac{1}{2} k(k-1) \\ &= \frac{1}{2} \left(\frac{(n-1)n(2n-1)}{6} - \frac{n(n-1)}{2} \right) \\ &= \frac{n(n-1)(n-2)}{6} = \binom{n}{3}. \end{aligned}$$

Hence, a normalized version of the total cophenetic index is $I_{\Phi, \text{norm}}(T) = I_{\Phi}(T) / \binom{n}{3}$. This normalized imbalance index is not minimal for all fully symmetric trees. For example, the cophenetic value of the two leftmost leaves of the fully symmetric tree in Figure 1b is two, and so both the un-normalized and normalized cophenetic indices of this tree will be nonzero.

Conservative Tree Balance Indices

Our axioms permit J to change discontinuously when we add rare types to the root. This is because Axioms 3 and 4 consider the addition of subtrees that have vanishingly small magnitude relative to other subtrees excluding their roots, whereas the relative size of the root of the entire tree is immaterial. For example, consider a two-node linear tree T in which the nonroot node has size δ , relative to the size of the root. Then $J(T) = 0$ by Axiom 4. But if we add another child to the root of T , also of relative size δ , then the J value of the new tree will be 1 (by Axiom 1), even as $\delta \rightarrow 0$. To make our index robust in such cases, we can add another axiom:

Axiom A.1 (Root limit). Let T be a tree with root r . Then, $J(T) \rightarrow 0$ as $S_r^*/S_r \rightarrow 1$.

But this new axiom conflicts with Axiom 1, which we must then modify, such that equal splits are no longer sufficient for maximal balance:

Axiom A.2 (Alternative maximum value). $J(T) \leq 1$ for all trees T , and $J(T) = 1$ only if T has equal splits. Furthermore, if T has equal splits and is a leafy tree then $J(T) = 1$.

We will call a tree balance index *conservative* if it conforms to these two alternative axioms in addition to Axioms 2, 3, 4, and 5. This name is appropriate because Axiom A.1 implies that a tree will be considered imbalanced unless there is strong evidence to the contrary (in the form of a relatively small root node). Every conservative index is both universal and robust.

One way to define a class of conservative indices is to add to Equation 1 a *nonroot dominance* factor $h: \mathbb{R}_{>0} \times \mathbb{R}_{>0} \rightarrow (0, 1]$ with $h(x_1, x_2) \rightarrow 0$ as $x_1/x_2 \rightarrow 0$, and $h(x_1, x_2) = 1$ if and only if $x_1 = x_2$. We then obtain

$$J := \frac{1}{\sum_{k \in \tilde{V}} g_k} \sum_{i \in \tilde{V}} g_i h_i W_i,$$

with $h_i = h(S_i^*, S_i)$. The role of h is to quantify the extent to which a node should be considered a leaf (which does not contribute to the index's value) as opposed to an internal node (which does). Adding this factor has no effect on the balance values assigned to leafy trees, including cladograms, because if an internal node i has zero size then $h_i = 1$. Setting $h(x_1, x_2) = x_1/x_2$, we can modify Equation 3 to obtain the specific conservative index

$$J^{1c} := \frac{1}{\sum_{k \in \tilde{V}} S_k^*} \sum_{i \in \tilde{V}} S_i^* \frac{S_i^*}{S_i} W_i^1.$$

We previously used J^1 instead of J^{1c} to denote the above index (Noble et al. 2022).

Alternative Axioms Proposed by Fischer et al. (2021)

Shortly after we posted a preprint version of the current article, Fischer et al. (2021) posted a preprint in which they proposed two alternative axioms for nonrobust, nonuniversal tree balance indices, such as Sackin's and Colless' indices. In these axioms, $\mathcal{BT}_n^*$ denotes the set of rooted bifurcating trees with n leaves, $\mathcal{T}_n^*$ is the set of all rooted trees with n leaves such that $d^+(i) > 1$ for all internal nodes i , and the tree balance index is denoted t .

Axiom A.3 (Fischer et al. minimum value). The caterpillar tree with n leaves is the unique tree minimizing t on $\mathcal{T}_n^*$ (if t is defined on multifurcating trees) or on $\mathcal{BT}_n^*$ (if t is defined only on bifurcating trees) for all $n \geq 1$.

Axiom A.4 (Fischer et al. maximum value). The fully symmetric bifurcating tree with n leaves is the unique tree maximizing t on $\mathcal{BT}_n^*$ for all $n = 2^h$ with $h \in \mathbb{N}_{\geq 0}$.

These axioms can be compared with our axioms if we consider only leafy trees in which all leaves have equal size (such as cladograms). Axiom A.4 is then just a special case of our more general Axiom 1 because the

fully symmetric bifurcating tree with n leaves is the only tree in BT_n^* that has equal splits. But Axiom A.3 is not necessarily consistent with our Axiom 2. In particular, as shown in Figure 4b, our index J^1 does not comply with Axiom A.3 in the case of multifurcating leafy trees. We can resolve this incompatibility with the following simplification:

Axiom A.5 (Alternative Fischer et al. minimum value). The caterpillar tree with n leaves is the unique tree minimizing t on BT_n^* for all $n \geq 1$ (whether or not t is defined on multifurcating trees).

J^1 is consistent with Axiom A.5 because, when we consider only bifurcating leafy trees in which all leaves have equal size, J^1 is equal to J_S (by Proposition 6), which is inversely proportional to I_S by definition, and the caterpillar tree is the unique bifurcating tree that maximizes I_S (Fischer et al. 2021). Although Axiom 1 does not necessarily imply Axiom A.5, it is reasonable to expect useful universal tree balance indices to satisfy both conditions.

Proof that the Index of Equation 1 Satisfies Our Five Axioms

Proof. **Axiom 1 (Maximum value):** We have $J \leq 1$ since g and W lie between zero and one by definition. Also if any internal node j of tree T does not split its descendants into at least two equal-magnitude subtrees then $W_j < 1$ by definition and so

$$\sum_{i \in \tilde{V}} g_i W_i < \sum_{i \in \tilde{V}} g_i \implies J(T) < 1.$$

Now, let T be a tree such that every internal node splits its descendants into at least two equal-magnitude subtrees. Then $W_i = 1$ for all $i \in \tilde{V}$ by definition. Hence,

$$J(T) = \frac{1}{\sum_{k \in \tilde{V}} g_k} \sum_{i \in \tilde{V}} g_i = 1.$$

Axiom 2 (Minimum value): We have $J \geq 0$ since g and W are always non-negative by definition. Also if T is a linear tree then $W_i = 0$ for all $i \in \tilde{V}$ by definition, and hence $J(T) = 0$. Conversely, if some internal node j has $d^+(j) > 1$ then $W_j > 0$ by definition and, because g_j must be positive by definition, we must have $J(T) > 0$.

Axiom 3 (Insensitivity): Adding a subtree to a leaf l changes the tree balance value via the contributions of two sets of nodes: the internal nodes of T_l (including l), and all other internal nodes. For each internal node, $i \in \tilde{V}(T_l)$, as $S_l^* / \sum_{j \in \tilde{V}(T)} S_j^* \rightarrow 0$ so also $S_i^* / \sum_{j \in \tilde{V}(T)} S_j^* \rightarrow 0$ (because $S_i^* \leq S_l^*$), which implies $g_i \rightarrow 0$ by definition, and hence all such contributions approach zero. The contribution of all other internal nodes also approaches zero because g and W are continuous by definition.

Axiom 4 (Linear limit): Let $i \in \tilde{V}(T)$ with $d^+(i) = 1$. Without loss of generality, let i_1 denote the original

child of i , and $i_2, \dots, i_p$ denote the newly added children of i . Adding subtrees to i changes the tree balance value via the contributions of the newly added nodes and of node i . As $S_{i_1} / S_i^* \rightarrow 1$, so $S_{i_k} / S_i^* \rightarrow 0$ for all $k \in \{2, \dots, p\}$. This implies that $S_{i_k} / \sum_{j \in \tilde{V}(T)} S_j^* \rightarrow 0$ and hence $g_{i_k} \rightarrow 0$ by definition for all $k \in \{2, \dots, p\}$. Therefore, the first contribution approaches zero. Also as $S_{i_1} / S_i^* \rightarrow 1$, we have $\max(S_{i_1} / S_i^*, \dots, S_{i_p} / S_i^*) \rightarrow 1$, and so $W_i \rightarrow 0$ by definition. Therefore, the second contribution also approaches zero.

Axiom 5 (Continuity): The continuity of J follows immediately from the continuity of g and W . $\square$

New Generalizations of Sackin's and Colless' Indices

The number of distinct subtrees that contain a given leaf l is equal to its number of ancestors, which is the same as v_l , the depth of l . Hence, Sackin's index is equivalent to the sum of the leaf counts of the subtrees rooted at each internal node. By extension, we can define a new, more general form of Sackin's index that accounts for node sizes:

$$I_{S, \text{gen}}(T) := \sum_{i \in \tilde{V}(T)} S_i^*,$$

where S_i^* is the magnitude of the subtree rooted at node i , excluding the root. In the special case of leafy trees in which all leaves have size one, we recover $I_{S, \text{gen}} = I_S$. This new index is not very useful for assessing tree balance because it increases with the total tree magnitude, but in our framework, it performs an important role as a normalizing factor.

If we let S_{i_1} denote the magnitude of the left branch of the subtree rooted at i , and S_{i_2} denote the magnitude of the right branch, then we can generalize Colless' index to account for node sizes in bifurcating trees:

$$I_{C, \text{gen}}(T) := \sum_{i \in \tilde{V}(T)} |S_{i_1} - S_{i_2}| = \sum_{i \in \tilde{V}(T)} S_i^* |p_{i_1} - p_{i_2}|,$$

where $p_{i_j} = S_{i_j} / S_i^*$. This definition reduces to I_C in the case of leafy trees in which all leaves have size one. The right-hand expression above clarifies that the contribution of each node to Colless' index is the product of the node's importance (i.e., its number of descendants) and its balance (the degree to which the node splits its descendants into two equal-magnitude subtrees). We further see that $I_{C, \text{gen}}(T) \leq I_{S, \text{gen}}(T)$ for all trees T (because $|p_{i_1} - p_{i_2}| \leq 1$ for all i_1, i_2), which suggests the normalization

$$I_{C, \text{gen}, \text{norm}} := \frac{I_{C, \text{gen}}}{I_{S, \text{gen}}} = \frac{1}{\sum_{k \in \tilde{V}} S_k^*} \sum_{i \in \tilde{V}(T)} S_i^* |p_{i_1} - p_{i_2}|.$$

This new generalization of Colless' index is more robust than the conventional form, in the sense that its value is insensitive to the addition or removal of relatively small nodes. $I_{C, \text{gen}, \text{norm}}$ also enables meaningful

comparison of trees with different numbers of leaves. But, the problem remains that $I_{C, \text{gen}, \text{norm}}$ applies only to bifurcating trees.

Other Balance Indices Based on Generalized Entropies

As defined by [Chao et al. \(2014\)](#), generalized entropies for $q \geq 0, q \neq 1$ are

$${}^qH := \frac{1}{q-1} \left(1 - \sum_{i=1}^P p_i^q \right).$$

Parameter q determines the sensitivity to the type frequencies. 0H is simply the richness (minus 1) of the population, which corresponds to ignoring the frequencies and just counting the types. For $0 < q < 1$, rare types are given more weight than implied by their proportion, whereas for $q > 1$ abundant types matter more. 2H is the Gini-Simpson coefficient. In the limit $q \rightarrow 1$, we recover the Shannon entropy 1H_e .

For $q > 0$, qH attains its maximum value if and only if all types have equal frequency $p_i = 1/m$:

$$\max({}^qH) = \frac{1}{q-1} \left(1 - \frac{1}{m^{q-1}} \right) = \frac{m^{q-1} - 1}{m^{q-1}(q-1)}.$$

We can therefore define a normalized balance score W_i^q for $q > 0, q \neq 1$ and $i \in \tilde{V}$:

$$W_i^q := \begin{cases} \frac{d^+(i)^{q-1}}{d^+(i)^{q-1} - 1} \left(1 - \sum_{j \in C(i)} p_{ij}^q \right) & \text{if } d^+(i) \geq 2 \\ 0 & \text{otherwise.} \end{cases}$$

Similarly, one can define W_i^q for $q > 0, q \neq 1$ based on the entropy defined by [Rényi \(1961\)](#):

$$W_i^q := \begin{cases} \frac{1}{(1-q)\log d^+(i)} \log \left(\sum_{j \in C(i)} p_{ij}^q \right) & \text{if } d^+(i) \geq 2 \\ 0 & \text{otherwise.} \end{cases}$$

In either case, a balance index J^q satisfying our axioms is

$$J^q := \frac{1}{\sum_{k \in \tilde{V}} S_k^*} \sum_{i \in \tilde{V}} S_i^* W_i^q,$$

for any $q > 0$. And in either case, $J^q \rightarrow J^1$ as $q \rightarrow 1$.

Proof of Proposition 6

Proof. By definition of J^1 , if T is a tree on n leaves with $d^+(i) = m > 1$ and $f(i) = 0$ for every internal node i then

$$J^1(T) = \frac{-1}{\sum_{k \in \tilde{V}} S_k} \sum_{i \in \tilde{V}} \sum_{j \in C(i)} S_j \log_m \frac{S_j}{S_i}.$$

The sum of subtree magnitudes over the set of all internal nodes is equal to the sum of v_i multiplied by leaf size over the set of all leaves:

$$I_{S, \text{gen}} := \sum_{k \in \tilde{V}} S_k = \sum_{k \in L} v_k f(k).$$

Summing first over the internal nodes and then over their children gives the same result:

$$\sum_{i \in \tilde{V}} \sum_{j \in C(i)} S_j = \sum_{i \in \tilde{V}} S_i = \sum_{i \in L} v_i f(i) = \sum_{i \in L} f(i) \sum_{j=1}^{v_i} 1.$$

Let $a(i, j)$ denote the ancestor of node i at distance j , with $a(i, 0) = i$ and $a(i, v_i) = r$ (the root) for all i . Then by extension,

$$\sum_{i \in \tilde{V}} \sum_{j \in C(i)} S_j \theta(S_i, S_j) = \sum_{i \in L} f(i) \sum_{j=1}^{v_i} \theta(S_{a(i, j)}, S_{a(i, j-1)}),$$

for any function θ . In particular, we have

$$\sum_{i \in \tilde{V}} \sum_{j \in C(i)} S_j \log_m \frac{S_j}{S_i} = \sum_{i \in L} f(i) \sum_{j=1}^{v_i} \log_m \frac{S_{a(i, j-1)}}{S_{a(i, j)}}.$$

Substituting this result into the expression for J^1 , we find

$$\begin{aligned} J^1(T) &= \frac{-1}{\sum_{k \in \tilde{V}} S_k} \sum_{i \in L} \sum_{j=1}^{v_i} f(i) \log_m \frac{S_{a(i, j-1)}}{S_{a(i, j)}} \\ &= \frac{-1}{\sum_{k \in \tilde{V}} S_k} \sum_{i \in L} f(i) \sum_{j=1}^{v_i} (\log_m S_{a(i, j-1)} - \log_m S_{a(i, j)}). \end{aligned}$$

The right-hand sum is a telescoping series that collapses to give

$$J^1(T) = \frac{-1}{\sum_{k \in \tilde{V}} S_k} \sum_{i \in L} f(i) (\log_m S_{a(i, 0)} - \log_m S_{a(i, v_i)}).$$

Now since i is a leaf, $\log_m S_{a(i, 0)} = \log_m S_i = \log_m f(i)$. Also $\log_m S_{a(i, v_i)} = \log_m S_r = \log_m S(T)$. Hence,

$$\begin{aligned} J^1(T) &= \frac{-1}{\sum_{k \in \tilde{V}} S_k} \sum_{i \in L} f(i) (\log_m f(i) - \log_m S(T)) \\ &= \frac{-1}{\sum_{k \in \tilde{V}} S_k} \sum_{i \in L} f(i) \log_m \frac{f(i)}{S(T)} \\ &= \frac{{}^1H_m(T) S(T)}{\sum_{k \in \tilde{V}} S_k} = \frac{{}^1H_m(T) S(T)}{I_{S, \text{gen}}(T)}. \end{aligned}$$

If additionally all leaves i of T have the same size $f(i)=f_0$ then $S(T)=nf_0$, ${}^1H_m(T)=\log_m n$, and $I_{S,\text{gen}}(T)=f_0 I_S(T)$, which implies $J^1(T)=n\log_m n/I_S(T)$. $\square$

Proof of Proposition 7

Proof. Since $\sum_{k \in \tilde{V}} n_k = I_S(T)$, the conditions are equivalent to

$$I_S(T)J(T) = \sum_{i \in \tilde{V}} n_i W_i = n \log_m n,$$

$$\text{with } W_i = W\left(\frac{n_{i_1}}{n_i}, \dots, \frac{n_{i_{p(i)}}}{n_i}\right),$$

where $n_{i_1}, \dots, n_{i_{p(i)}}$ are the children of i . Let T be a tree in $\mathcal{T}_{n,m}^*$ and i be an internal node of T . Then, $T_i \in \mathcal{T}_{n_i,m}^*$ and $T_j \in \mathcal{T}_{n_j,m}^*$ for every child j of i . Therefore

$$I_S(T_i)J(T_i) = n_i W_i + \sum_{j \in C(i)} J(T_j) = n_i W_i + \sum_{j \in C(i)} n_j \log_m n_j.$$

Also, $I_S(T_i)J(T_i) = n_i \log_m n_i$, so we have

$$n_i W_i + \sum_{j \in C(i)} n_j \log_m n_j = n_i \log_m n_i$$

$$\implies W_i = \log_m n_i - \sum_{j \in C(i)} \frac{n_j}{n_i} \log_m n_j.$$

Since $\sum_{j \in C(i)} n_j = n_i$, this implies

$$\begin{aligned} W_i &= \sum_{k \in C(i)} \frac{n_k}{n_i} \log_m n_i \\ &\quad - \sum_{j \in C(i)} \frac{n_j}{n_i} \log_m n_j = - \sum_{j \in C(i)} \frac{n_j}{n_i} \log_m \frac{n_j}{n_i} = W_i^1. \quad \square \end{aligned}$$

Proof that J_S is a Sound Tree Balance Index

Proof. By the definition of Mir et al. (2018), a sound tree balance index J is such that $J(T)$ is maximal if and only if T is fully symmetric. The fully symmetric full m -ary tree on n leaves is the unique tree that minimizes I_S among full m -ary trees on n leaves. This minimum value is $\min_{n,m} I_S = n \log_m n$ (since every leaf l has the same depth $v_l = \log_m n$). Because $J_S := n \log_m n / I_S$ is defined only on full m -ary trees, it follows that $J_S(T)$ is maximal if and only if T is fully symmetric. $\square$