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Statistics and Partitioning of Species Diversity, and Similarity among Multiple Communities

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MINI- REVIEW

Minireviews provides an opportunity to summarize existing knowledge of selected ecological areas, with special emphasis on current topics where rapid and significant advances are occurring. Reviews should be concise and not too wide-ranging. All key references should be cited. A summary is required.

Statistics and partitioning of species diversity, and similarity among multiple communities

Russell Lande

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Species richness, Shannon information, and Simpson diversity are the three most commonly used nonparametric measures of species diversity. The sampling bias and variance of these measures differ greatly. Species richness may be seriously underestimated for even very large samples from a speciose community. The bias in species richness and Shannon information depend on unknown parameters of the species abundance distribution. An unbiased estimator exists only for Simpson diversity. Each of these measures is concave, so that the total diversity in a pooled set of communities exceeds (or equals) the average diversity within communities. The total diversity in a set of communities can therefore be partitioned into positive, additive components within and among communities, corresponding to α - and β -diversity. Partitioning Simpson diversity corresponds to an analysis of variance. The proportion of the total diversity found within communities provides a natural measure of similarity among multiple communities. The expected similarity among multiple random samples from the same community depends on the number of samples and on the underlying measure of diversity.

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Measures of species diversity play a central role in ecology and conservation biology (Whittaker 1960, 1972, Williams 1964, MacArthur 1965, Peet 1974, Pielou 1975, Grasse et al. 1979, Magurran 1988, Noss and Cooperrider 1994). The most commonly employed measures of species diversity are species richness (number of species present in a community), and those based on species frequencies involving Shannon information, H , and Simpson concentration, λ .

Whittaker (1960, 1972) defined the important concepts of species diversity within and among communities (α - and β -diversity), and the total species diversity in a set of communities (γ -diversity). Various measures

of species diversity among communities have been proposed, especially for patterns of species richness along environmental gradients (Whittaker 1960, 1972, MacArthur 1965, 1972, Pielou 1975, Allan 1975, Routledge 1977, Wilson and Mohler 1983, Wilson and Shmida 1984, Magurran 1988).

A measure of species diversity should ideally be *nonparametric* and *statistically accurate*. It should be applicable to any community independent of species abundance distribution, and should have small bias and sampling variance in samples of moderate size. An important property for a diversity measure, first discussed by Lewontin (1972) for genetic diversity, is *strict*

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concavity. This means that the total diversity in a pooled set of communities equals or exceeds the average diversity within communities, with equality only for identical communities. Lewontin partitioned the total diversity into a sum of the average diversity within communities and the diversity among communities. As we will see below, such an *additive partition* of diversity leads naturally to a measure of similarity among multiple communities.

Although considerable effort has been devoted to analyzing basic statistics of diversity measures, less attention has been given to theoretical or empirical partitioning of total species diversity within and among communities, and some of the most useful results are scattered through the genetic, statistical and ecological literature. Here I collect and extend results on the three most popular measures of species diversity, and critically evaluate their relative merits with respect to the above properties. Finally I discuss the relative merits of these measures for assessing similarity among communities and for extrapolating total species diversity in a region from samples within and among communities.

Nonparametric measures of species diversity

The most popular measures of species diversity are nonparametric and do not depend on any particular species abundance distribution, such as the log series (Fisher et al. 1943), broken stick (MacArthur 1957), or lognormal (Preston 1948) models, from which real communities will deviate to some extent.

Species richness

This is simply the number of species in a community (or sample) based on presence, rather than relative abundance.

Shannon information

Let p_i be the frequency of species i in a community. The average information per individual is

$$H = - \sum_{i=1}^S p_i \ln p_i \quad (1)$$

Shannon and Weaver (1962). For a given number of species, S , the information reaches its maximum value, $\ln S$, when all species are equally frequent in the community. Many authors use the exponential of Shannon information, e^H , which for a given number of species has a maximum equal to S .

Simpson concentration

The probability that two randomly chosen individuals from a given community are the same species, called the "concentration" by Simpson (1949), is

$$\lambda = \sum_{i=1}^S p_i^2. \quad (2)$$

$1 - \lambda$ is the probability that two randomly chosen

individuals from a given community are different species, also called the Gini coefficient, which can be used as a measure of diversity (Pielou 1969). The inverse of Simpson concentration, $1/\lambda$, is often employed to measure species diversity, and for a given number of species, S , in a community it has a maximum value equal to S when all species are equally frequent.

Statistics of species diversity

If two or more samples are known to be different, either a priori because they come from different communities or after rejection of the null hypothesis of homogeneity among samples, it may then be appropriate to test the hypothesis that one community is more diverse than another, using some measure of species diversity. Hutcheson (1970) described a t -test for the significance of differences in the Shannon information measure of diversity in large samples. More generally, a resampling scheme such as the jackknife or bootstrap is suitable (Efron 1982, Magurran 1988). To assess the accuracy of commonly used nonparametric measures of species diversity, I here collect and derive approximate formulas for their sampling bias and variance.

Species richness

Let S be the actual number of species in a community composed of a very large (effectively infinite) number of individuals. The frequency of the i th species is p_i . Using a carat to denote a sample value, the number of species in a sample of size N is \hat{S} , with mean

$$E[\hat{S}] = S - \sum_{i=1}^S (1 - p_i)^N \quad (3a)$$

(Grassle and Smith 1976). The rarefaction formula of Hurlbert (1971) gives the analog of (3a) for samples from a finite community. Species i is likely to be present in a sample of size N only if $p_i N > 1$. Hence in a highly diverse community the observed number of species, \hat{S} , may greatly underestimate the actual number of species, S , because rare species frequently will be absent from even very large samples. Several methods have been developed for estimating the actual number of species by extrapolation from samples of various sizes (reviewed in Colwell and Coddington 1994). All extrapolation methods for estimating the actual number of species in a community make implicit or explicit assumptions about the form of the abundance distribution of rare species, and hence may be in error to an unknown extent.

The sampling variance of species richness is exactly (corrected from Strömrgren et al. 1973)

$$\begin{aligned} \text{Var}[\hat{S}] = & \sum_{i=1}^S (1 - p_i)^N [1 - (1 - p_i)^N] \\ & + 2 \sum_{i>j} [(1 - p_i - p_j)^N - (1 - p_i)^N (1 - p_j)^N]. \end{aligned} \quad (3b)$$

The first (single) summation contains the variances of presence (1) versus absence (0) of each species. The second (double) summation contains the covariances of presence versus absence for pairs of species, which are negative and become negligible in comparison to the variances for sufficiently large sample sizes.

Shannon information

In a random sample of N individuals from a community, the diversity calculated using the estimated species frequencies \hat{p}_i , is denoted as \hat{H} . For large N this has approximate mean and variance

$$E[\hat{H}] \cong H - \frac{S-1}{2N} \quad (4a)$$

$$\text{Var}[\hat{H}] \cong \left[\sum_{i=1}^S p_i (\ln p_i)^2 - H^2 \right] / N \quad (4b)$$

(Pielou 1966, Hutcheson 1970, Bowman et al. 1971). Note that the bias in \hat{H} depends on the actual number of species, S , which is generally unknown. Hence an unbiased estimator of Shannon information does not exist. Accurate estimation of H requires sampling large numbers of individuals, with $2N$ much greater than the actual number of species in the community.

Simpson diversity

Pielou (1975) and Patil and Taillie (1982) draw an analogy between diversity which measures the variety of categorical (species) identities, and variance which measures the dispersion in quantitative measurements. It does not appear to have been previously noticed that the Simpson measure of species diversity within a community, $1 - \lambda$, can be expressed precisely as a variance. If individual k of species i is denoted as a point in S -dimensional space, with coordinates (x_{1k}, \dots, x_{Sk}) where $x_{ik} = 1$ and all other coordinates are 0, then the total variance per individual in species identity within the community is

$$\begin{aligned} \sum_i \sum_k E[(x_{ik} - p_i)^2] &= \sum_i p_i (1 - p_i) \\ &= 1 - \lambda. \end{aligned} \quad (5a)$$

Because this is a variance it follows directly that in a random sample of N individuals from a community, estimates of the Simpson diversity, $1 - \hat{\lambda}$, calculated using the estimated species frequencies \hat{p}_i , have mean exactly

$$E[1 - \hat{\lambda}] = \left(1 - \frac{1}{N}\right)(1 - \lambda). \quad (5b)$$

Thus an unbiased estimator of Simpson diversity is

$$1 - \tilde{\lambda} = \left(\frac{N}{N-1}\right)(1 - \hat{\lambda}). \quad (5c)$$

For large samples the approximate variance of Simpson diversity is

$$\text{Var}[1 - \hat{\lambda}] \cong \frac{4}{N} \left[\sum_{i=1}^S p_i^3 - \lambda^2 \right]. \quad (5d)$$

which is the same as that for Simpson's (1949) unbiased estimator of concentration, $\tilde{\lambda}$.

Fig. 1 compares the sampling properties of these three diversity measures for communities of low or high diversity. It can be seen that the estimated species richness can greatly underestimate the actual number of species even for very large samples in a highly diverse community. The standard deviation of species richness also is rather large. Shannon information has a substantial bias for small samples, but the bias is small when $2N \gg S$. The standard deviation of Shannon diversity is much smaller than that for species richness. Simpson diversity, $1 - \tilde{\lambda}$ is not only unbiased, but also has the smallest standard deviation.

Concavity of species diversity measures

A desirable property of a measure of species diversity is that the total diversity in a set of communities should be greater than or equal to the (weighted) average diversity within the communities (Lewontin 1972). Let D_T be the total species diversity in a set of communities (or samples), computed using the weighted average species frequencies among communities. Denote the diversity within community j as D_j , and let the proportional "weight" of this community be q_j such that $\sum_j q_j = 1$. The weights associated with each community may reflect the relative abundance of the communities, the sizes of samples from the communities, or equal weights. A diversity measure is *strictly concave* when

$$D_T \geq \sum_j q_j D_j \text{ with equality only for identical communities.} \quad (6)$$

Species richness obviously is strictly concave.

A continuous measure of diversity based on species frequencies is strictly concave if and only if its matrix of second derivatives is negative definite (Marcus and Minc 1964).

The Shannon information measure of species diversity, H , is strictly concave (Aczél and Daroczy 1975: 35).

The Simpson concentration, λ , is strictly *convex*. Hence the diversity measure $1 - \lambda$ is strictly concave. However, the more commonly used inverse Simpson measure of diversity, $1/\lambda$, is in general not concave (Patil and Taillie 1982). Thus in some cases using the inverse Simpson measure, the total diversity in a set of communities may be less than the average diversity

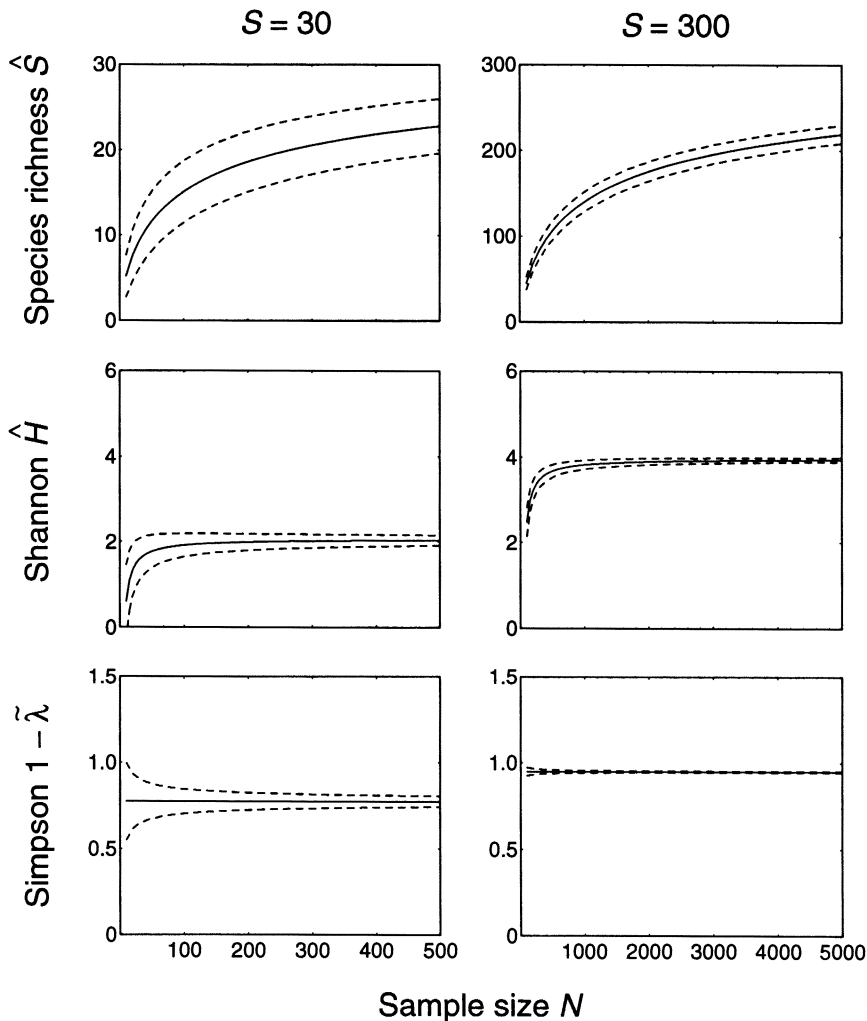


Fig. 1. Estimated species richness, \hat{S} , Shannon diversity, \hat{H} , and Simpson diversity, $1 - \hat{\lambda}$, as a function of sample size, N , for hypothetical communities in which the species abundance distribution is lognormal with the standard deviation of the natural logarithm of species frequencies $\sigma = 2$. Solid lines give the mean diversity, and dashed lines show approximate 95% confidence intervals (plus and minus two standard deviations). In communities with a total of $S = 30$ or 300 species the frequency of the most abundant species is 42.6% or 17.4%, respectively.

within communities, which implies the possibility of a negative diversity among communities. The simplest case is that of communities with only two species in which the frequency of the most common species exceeds $(1 + \sqrt{3})/2 \cong 0.7887$. The same result can occur also in more diverse communities with substantial unevenness in species frequencies (see Table 1).

Additive partition of diversity

The total species diversity in a pooled set of communities can be partitioned into additive components within and among communities, so that total diversity and its components have the same units and can be compared directly. An additive partition of diversity therefore

seems more natural than the multiplicative partitions described by Whittaker (1960, 1972) and Routledge (1979). A partition would be most easily interpreted if the different components of diversity could be expressed using the same general formula.

Consider a set of communities in which the frequency of species i in community j is p_{ij} , such that $\sum_i p_{ij} = 1$. Let q_j be the proportional weight associated with community j , based on its sample size or importance. The total species diversity, D_T , is defined in terms of the weighted average frequency of species i in the pooled set of communities, $\bar{p}_i = \sum_j q_j p_{ij}$. For a measure of diversity that is concave, the total diversity in a set of communities is always greater than or equal to the weighted average diversity within communities (eq. 6), and can therefore be additively partitioned into non-negative components within and among communities,

Table 1. Examples in which the inverse Simpson measure of species diversity, $1/\lambda$, violates concavity, so that the diversity of a pooled set of communities is less than the average diversity within communities. Pooled communities have equal weights.

	Example 1			Example 2		
	Species frequencies in community			Species frequencies in community		
	1	2	pooled	1	2	pooled
$p_1 =$	0.8	1.0	0.9	0.50	0.3529	0.42645
$p_2 =$	0.2	0.0	0.1	0.26	0.2022	0.23110
$p_3 =$				0.09	0.0955	0.09275
$p_4 =$				0.05	0.0703	0.06015
$p_5 =$				0.04	0.0641	0.05205
$p_6 =$				0.03	0.0578	0.04390
$p_7 =$				0.02	0.0515	0.03575
$p_8 =$				0.01	0.1057	0.05785
$1/\lambda =$	1.4706	1.0000	1.2195	3.019	4.981	3.895
	$(1/\lambda_1 + 1/\lambda_2)/2 = 1.2353$			$(1/\lambda_1 + 1/\lambda_2)/2 = 4.000$		

$$D_T = D_{\text{among}} + \bar{D}_{\text{within}} \quad (7)$$

where $\bar{D}_{\text{within}} = \sum_j q_j D_j$.

Species richness

In a pooled set of communities the total species richness is S_T and the species richness in community j is S_j . The among-community component of total species richness is

$$S_T - \bar{S}_{\text{within}} = \sum_j q_j (S_T - S_j). \quad (8)$$

$S_T - S_j$ is the discrepancy between total species richness and that in the j th community, which is non-negative. Formula (8) can be compared with Whittaker's multiplicative partition of total diversity in which β -diversity is measured by the ratio $\beta = S_T / \bar{S}_{\text{within}}$.

Shannon information

The information diversity among communities can be expressed as

$$H_{\text{among}} = -\sum_i \bar{p}_i \ln \bar{p}_i - \sum_j q_j H_j \quad (9a)$$

$$= \sum_j q_j H_{(j,T)}. \quad (9b)$$

$$H_{(j,T)} = \sum_i p_{ij} \ln(p_{ij} / \bar{p}_i) \quad (9c)$$

is the "discrimination information" between community j and the pooled set of communities, which is non-negative (Kullback 1959, Rényi 1961, Aczél and Daroczy 1975). MacArthur (1965) and MacArthur and Wilson (1967) employed (9a) with equal weights to measure the component of bird species diversity between two communities. Lewontin (1972) used (9a) to partition genetic diversity in human populations, and his method was elaborated for species diversity by Allan (1975). In the special case when no species is

present in more than a single community the diversity among communities is maximized and takes the usual form for information

$$\max H_{\text{among}} = -\sum_j q_j \ln q_j.$$

This occurs, for example, in the "hierarchical model" of Pielou (1969, 1975) where different communities are composed of distinct taxonomic groups.

Simpson diversity

For the measure of species diversity, $D = 1 - \lambda$, based on Simpson's measure of concentration, the diversity among communities can be expressed in terms of the variance in species frequencies among communities. Weighting the j th community by its overall frequency or importance, q_j , the variance among communities is

$$\begin{aligned} d^2 &= \sum_j q_j \sum_i (p_{ij} - \bar{p}_i)^2 \\ &= \sum_j q_j \lambda_j - \sum_i \bar{p}_i^2. \end{aligned} \quad (10a)$$

Therefore

$$\begin{aligned} D_T &= 1 - \sum_i \bar{p}_i^2 \\ &= d^2 + \bar{D}_{\text{within}}. \end{aligned} \quad (10b)$$

d^2 is thus a natural measure of species diversity among communities. $\sum_i (p_{ij} - \bar{p}_i)^2$ is the squared distance in species frequencies between community j and the pooled set of communities, which is non-negative.

Essentially the same approach to partitioning *genetic* diversity within and among populations was developed by Nei (1973, 1987). Patil and Taillie (1982) noted its applicability to partitioning species diversity at two or more levels, e.g. sets of communities from different biotic provinces. As already shown in formula (5a), the Simpson diversity within a community is also a variance. Thus, using multivariate analysis of variance, the

total species diversity in a set of communities can be partitioned into additive components of the same functional form. This approach facilitates statistical analysis of components of species diversity. Statistical tests on components of variance based on categorical variables can be performed using methods in (Searle et al. 1992) or (Efron 1982).

Similarity among multiple communities

Employing a partition of total species diversity into additive components within and among multiple communities, a natural measure of community similarity is

$$\psi_D = \bar{D}_{\text{within}}/D_T$$

$$= 1 - D_{\text{among}}/D_T \quad (11)$$

which ranges between 0 and 1.

For species richness, this reveals that Whittaker's measure of β -diversity, $S_T/\bar{S}_{\text{within}}$, is not actually a diversity, but rather the inverse of community similarity in species composition.

Based on the Simpson measure of genetic diversity (heterozygosity) Nei (1973, 1987) developed an analogous measure of *genetic similarity* among populations (G_{ST}) which has been widely used to describe the genetic structure of populations. To measure community similarity based on Simpson diversity we must use $1 - \hat{\lambda}$ to estimate diversity within samples in order that the similarity measure not exceed unity, which it could when using the unbiased estimator $1 - \tilde{\lambda}$ for a finite number of samples (as for the Morisita index of similarity between two communities based on the unbiased Simpson concentration [Morisita 1959, Horn 1966, Wolda 1981]).

Standardized measures of community similarity, ranging from 0 to 1, generally are biased downward, so that the true similarity between communities, estimated from random samples, tends to be underestimated. This can be seen most clearly in the expected similarity among random samples from the same community.

Consider first the simplest situation in which a very large (effectively infinite) number of samples of the same size, N , are taken from the same (infinitely large) community. Then the total diversity of the pooled samples is the same as that in the community, D_T , and the similarity among communities is

$$\psi_{\hat{D}} = E[\hat{D}]/D_T \quad (12)$$

in which $E[\hat{D}]$ is the mean value of any one of the three diversity measures in formulas (3a), (4a) and (5b).

Fig. 2 plots these expected similarities among multiple samples from the same community. The amount of bias in the similarity measures parallels that for the corresponding diversity measures. The similarity in species richness among replicate samples has a large bias,

because different samples contain different sets of the rarest species. The bias in similarity of Shannon information among replicate samples depends on the ratio of total number of species in the community to $2N$ times the diversity in the community. The bias in similarity based on Simpson concentration depends only on the sample size and is small for moderately large samples.

We also can examine the ratio of the expected diversity within random samples of size N from the same community to the expected total diversity in n random samples of size N from the same community, $E[\hat{D}(N)]/E[\hat{D}(nN)]$. For an infinite number of samples, $n = \infty$, this equals the similarity coefficient in formula (12). For a finite number of samples, $n < \infty$, this ratio indicates the approximate behavior of the general formula (11) applied to the similarity among multiple samples from the same community (see Fig. 2).

Discussion

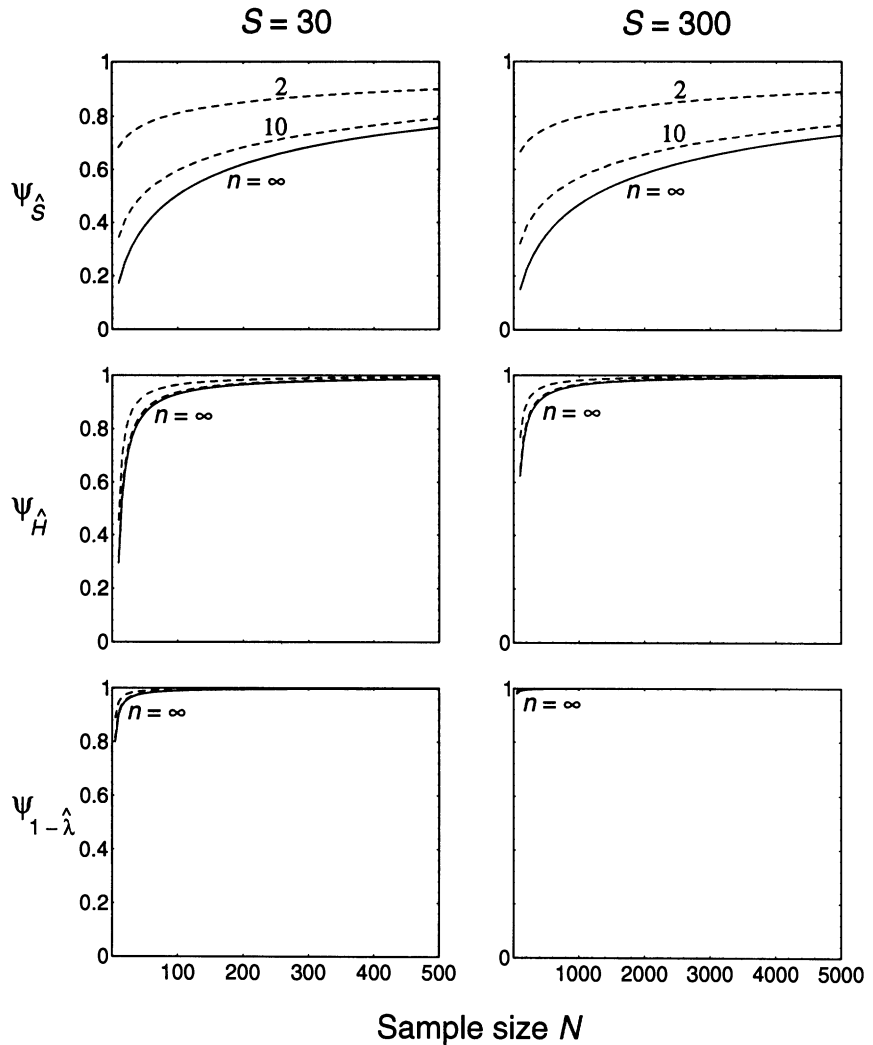
Species richness is the most widely used measure of diversity, because of its simplicity in data acquisition and analysis. However, it has a well-known statistical weakness of a potentially large sampling bias, in that rare species often will be absent even in large samples or exhaustive surveys.

The Shannon information measure of species diversity, though popular, has a rather tenuous foundation in ecological theory, as noted by Pielou (1966, 1969), Hurlbert (1971), and May (1975). In samples from speciose communities, the Shannon diversity may have a substantial bias (Hutcheson 1970, Bowman et al. 1971). For both species richness and Shannon information measures of diversity, the bias depends on the actual number of species in a community, which generally is unknown, so that an unbiased estimator of species richness or Shannon information does not exist.

May (1975) preferred diversity measures based on the Simpson index to those based on information, because of the apparent relationship of Simpson's index to variances. However, the commonly used inverse Simpson diversity $1/\lambda$ is not concave. When partitioning human genetic diversity, Lewontin (1972) noted that a diversity measure should have the property that the total diversity in a pooled set of communities is greater than or equal to the average diversity within communities. Violation of this property of concavity can produce the uninterpretable result of a negative diversity among communities.

In contrast with the inverse Simpson diversity, the Simpson diversity measure $1 - \lambda$, which is the probability that two randomly chosen individuals are different species, has a number of advantages. $1 - \lambda$ is precisely the variance of species identity within a community, for which an unbiased estimator exists. Using this measure the total diversity in a set of communities can be

Fig. 2. Expected similarity among n samples of N individuals from the same community, based on partition of three measures of the total species diversity in the samples. The number and abundance distribution of species in the community are as in Fig. 1. Solid lines give the exact proportion of the total species diversity in the community expected within samples. Dashed lines are approximations, based on the ratio of the expected diversity within samples to the expected total diversity in $n = 2$ or 10 samples.



partitioned into the average diversity within the communities plus the diversity among communities, with the latter equal to the variance in species frequencies among communities. This measure of species diversity thus facilitates statistical partitioning of species diversity within and among communities using analysis of variance.

Partition of total species diversity in a set of communities into additive components within and among communities provides a unifying framework with which to measure diversity at different levels of organization using the same general formula, so that α -, β -, and γ -diversity are measured in the same way. Additive partition of diversity also suggests a natural measure of similarity among multiple communities, the proportion of the total diversity found within communities.

Coefficients of community similarity inherit the statistical sampling properties of the diversity measures

on which they are based. The similarity among a set of communities can be expected to decrease with increasing number of communities in the set because the total diversity increases as more distinct communities are included. Sampling effects also contribute to decreasing the expected similarity among communities, because random samples deviate to some extent from the actual communities.

The expected similarity among random samples from the same community decreases with increasing number of samples because more samples contain a greater total diversity, more closely approximating that in the actual community. For species richness, the expected similarity among random samples from the same community can be substantially less than unity, even with very large sample sizes. For Shannon information this downward bias in community similarity becomes small in samples of moderate size. For Simpson diversity

$1 - \lambda$ the expected similarity among random samples from the same community closely approaches unity for modest sample sizes.

Measures of community similarity should not be employed to test the null hypothesis that different samples are drawn from the same community. When random samples are available, analysis of contingency tables (containing counts of individuals per species in different samples) using Chi-squared or likelihood (G) statistics provides a more general and powerful method for detecting heterogeneity in species abundance distributions among communities (Gokhale and Kullback 1978, Fienberg 1981, Sokal and Rohlf 1995).

The measures of similarity among multiple communities proposed here provide a natural description of community structure based on additive partition of species diversity within and among communities, which can be readily extended to both higher and lower levels of organization, e.g. multiple samples within communities, and multiple communities within geographic provinces or biomes.

Estimating the total species diversity for some taxonomic group in a region generally requires two kinds of extrapolations: (1) using a sample from a community to estimate species diversity within the community, and (2) using a sample of communities to estimate total species diversity within a region. Estimating species richness from even a large sample of a speciose community may require (explicit or implicit) assumption of a particular form of species abundance distribution to extrapolate the number of rare species (see Colwell and Coddington 1994). Accurate estimates of Shannon information diversity can be achieved for moderate sample sizes by assuming a form of species abundance distribution, or more generally by using sample sizes much larger than the actual number of species in the community. For the Simpson diversity the unbiased estimator (eq. 5c) accurately extrapolates the species diversity in a community even from a sample of modest size.

Estimation of total species diversity in a region, based on randomly chosen communities, can be accomplished by partitioning observed species diversity within and among communities, and extrapolating to the actual number of communities in the region. Sampling effects will cause a fraction of the average diversity within communities to appear as diversity among communities. Adjustments for this can be made by having multiple samples from each community, or by using the expected similarity among different samples from the same community (Fig. 2). Accurate estimation of total species diversity in a region based on a hierarchical random sampling scheme is most easily accomplished using an analysis of variance for Simpson diversity.

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References

- Aczél, J. and Daroczy, Z. 1975. On measures of information and their characterizations. – Academic Press, New York.
- Allan, J. D. 1975. Components of diversity. – *Oecologia* 18: 359–367.
- Bowman, K. O., Hutcheson, K., Odum, E. P. and Shenton, L. R. 1971. Comments on the distribution of indices of diversity. – In: Patil, G. P., Pielou, E. C. and Walters, W. E. (eds), *Statistical ecology*. Volume 3. Many species populations, ecosystems, and systems analysis. Pennsylvania State Univ. Press, University Park, PA, pp. 315–366.
- Colwell, R. K. and Coddington, J. A. 1994. Estimating terrestrial biodiversity through extrapolation. – *Philos. Trans. R. Soc. Lond. B* 345: 101–118.
- Efron, B. 1982. The jackknife, the bootstrap and other resampling plans. – Society for Industrial and Applied Mathematics, Philadelphia, PA.
- Fienberg, S. E. 1981. The analysis of cross-classified categorical data. 2nd edn. – MIT Press, Cambridge, MA.
- Fisher, R. A., Corbet, A. S. and Williams, C. B. 1943. The relation between the number of species and the number of individuals in a random sample of an animal population. – *J. Anim. Ecol.* 12: 42–58.
- Gokhale, D. V. and Kullback, S. 1978. The information in contingency tables. – Marcel Dekker, New York.
- Grassle, J. F. and Smith, W. 1976. A similarity measure sensitive to the contribution of rare species and its use in investigation of variation in marine benthic communities. – *Oecologia* 25: 13–22.
- , Patil, G. P., Smith, W. and Taillie, C. 1979. Ecological diversity in theory and practice. – International Co-operative Publishing House, Fairland, MD.
- Horn, H. S. 1966. Measurement of “overlap” in comparative ecological studies. – *Am. Nat.* 100: 419–424.
- Hurlbert, S. H. 1971. The nonconcept of species diversity: a critique and alternative parameters. – *Ecology* 52: 577–586.
- Hutcheson, K. 1970. A test for comparing diversities based on the Shannon formula. – *J. Theor. Biol.* 29: 151–154.
- Kullback, S. 1959. Information theory and statistics. – Wiley, New York.
- Lewontin, R. C. 1972. The apportionment of human diversity. – *Evol. Biol.* 6: 381–398.
- MacArthur, R. H. 1957. On the relative abundance of bird species. – *Proc. Natl. Acad. Sci. USA* 43: 293–295.
- 1965. Patterns of species diversity. – *Biol. Rev.* 40: 510–533.
- 1972. *Geographical ecology*. – Harper & Row, New York.
- and Wilson, E. O. 1967. The theory of island biogeography. – Princeton Univ. Press, Princeton, NJ.
- Magurran, A. E. 1988. Ecological diversity and its measurement. – Princeton Univ. Press, Princeton, NJ.
- Marcus, M. and Minc, H. 1964. A survey of matrix theory and matrix inequalities. – Dover, New York.
- May, R. M. 1975. Patterns of species abundance and diversity. – In: Cody, M. L. and Diamond, J. M. (eds), *Ecology and evolution of communities*. Harvard Univ. Press, Cambridge, MA, pp. 81–120.
- Morisita, M. 1959. Measuring of interspecific association and similarity between communities. – *Mem. Fac. Sci. Kyushu Univ. Ser. E Biol.* 3: 65–80.
- Nei, M. 1973. Analysis of gene diversity in subdivided populations. – *Proc. Natl. Acad. Sci. USA* 70: 3321–3323.
- 1987. *Molecular evolutionary genetics*. – Columbia Univ. Press, New York.
- Noss, R. F. and Cooperrider, A. Y. 1994. Saving nature's legacy, protecting and restoring biodiversity. – Island Press, Washington.

- Patil, G. P. and Taillie, C. 1982. Diversity as a concept and its measurement. – *J. Am. Stat. Assoc.* 77: 548–561.
- Peet, R. K. 1974. The measurement of species diversity. – *Annu. Rev. Ecol. Syst.* 5: 285–307.
- Pielou, E. C. 1966. Shannon's formula as a measure of specific diversity: its use and misuse. – *Am. Nat.* 100: 463–465.
- 1969. *An introduction to mathematical ecology.* – Wiley, New York.
- 1975. *Ecological diversity.* – Wiley, New York.
- Preston, F. W. 1948. The commonness and rarity of species. – *Ecology* 29: 254–283.
- Rényi, A. 1961. On measures of entropy and information. – In: Neyman, J. (ed.), *Proceedings of the 4th Berkeley Symposium on Mathematical Statistics and Probability*, Vol. 1. Univ. of California Press, Berkeley, CA, pp. 547–561.
- Routledge, R. D. 1977. On Whittaker's components of diversity. – *Ecology* 58: 1120–1127.
- 1979. Diversity indices: which ones are admissible. – *J. Theor. Biol.* 76: 503–515.
- Searle, S. R., Casella, G. and McCulloch, C. E. 1992. *Variance components.* – Wiley, New York.
- Shannon, C. E. and Weaver, W. 1962. *The mathematical theory of communication.* – Univ. of Illinois Press, Urbana, IL.
- Simpson, E. H. 1949. Measurement of diversity. – *Nature* 163: 688.
- Sokal, R. R. and F. J. Rohlf. 1995. *Biometry. The principles and practice of statistics in biological research*, 3rd ed. – Freeman, San Francisco.
- Strömberg, T., Lande, R. and Engen, S. 1973. Intertidal distribution of the fauna on muddy beaches in the Borgefjord area. – *Sarsia* 53: 49–70.
- Whittaker, R. H. 1960. Vegetation of the Siskiyou Mountains, Oregon and California. – *Ecol. Monogr.* 30: 279–338.
- 1972. Evolution and measurement of species diversity. – *Taxon* 21: 213–251.
- Williams, C. B. 1964. *Patterns in the balance of nature and related problems in quantitative ecology.* – Academic Press, London.
- Wilson, M. V. and Mohler, C. L. 1983. Measuring compositional change along gradients. – *Vegetatio* 54: 129–141.
- and Shmida, A. 1984. Measuring beta diversity with presence-absence data. – *J. Ecol.* 72: 1055–1064.
- Wolda, H. 1981. Similarity indices, sample size and diversity. – *Oecologia* 50: 296–302.