JACKKNIFING AND BOOTSTRAPPING SOME INDICES OF PATTERN DETECTION

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ABSTRACT

A comparison of the average degree of nonrandomness of patchiness in two or more populations as well as the degree of their statistical accuracy is often desired. Both Jackknife and Bootstrap provide nonparametric estimation of standard error of an estimator and thereby making possible the significance tests and the establishment of confidence intervals and testing the significance. In this paper we examine the statistical accuracy (standard error, bias, mean squared error) and provide methods for estimating confidence intervals for three indices of spatial dispersion pattern.

Key-words: Statistical methods, Jackkniefing, Bootstrapping, pattern detection

INTRODUCTION

Techniques of pattern detection abound in ecological literature (Greig-Smith, 1983; Ludurig and Reynolds, 1988; Upton and Fingleton, 1985). Some methods use statistical distributions of indices of dispersion for detecting and measuring spatial pattern of species populations. An obvious requirement of a good measure of dispersion is the relative insensitiveness to changes in density (Pielou, 1978). Morisita (1959, 1971) developed an index of dispersion \( I_\delta \) that is insensitive to changes in density caused by random thinning. This index which is based on a diversity measure proposed by Simpson (1949) is as follows:

\[
I_\delta = \frac{Q \sum_{i=1} x_i (x_i - 1)}{N (N - 1)}
\]

where \( x_i \) equals the number of individuals in the \( i \)th quadrat (\( i = 1, \ldots, Q \)) \( \sum x_i - N \) while \( N \) equals the sum of \( x_i \).

Lloyd (1967) proposed an ‘index of mean crowding’ and an index of patchiness. Mean crowding is defined as:

\[
\bar{m} = \frac{1}{N} \sum_{i=1} x_i (x_i - 1)
\]

The degree of crowding as measured by \( \bar{m} \) is dependent on the degree of clumping and the population density. However, Lloyds’ index patchiness \( C \) which is scaled to cancel out the population density effect out of the measurement is unaffected when some members of the population are removed at random.

\[
C = \frac{\bar{m}}{\lambda}
\]

where \( \lambda \) is the mean density per quadrat.

A comparison of the average degree of nonrandomness of patchiness in two or more populations as well as the degree of their statistical accuracy is often desired. Both jackknife and bootstrap provide nonparametric estimation of standard error of an estimator and thereby making possible the significance tests and the establishment of confidence intervals and testing the significance. These computer-intensive technique also peranist bias reduction. The jackknife method (Tukey, 1958) has been applied to ecological diversity indices (Zahl, 1977; Adams and McCune, 1979; Heltshe and Forrestea, 1985), population size estimation (Burnham and Overton, 1979), genetic distance estimates (Mueller, 1979) measures of niche overlap (Mueller and Altenberg, 1985) and similarity index (Smith et al., 1986).

The bootstrap method (Efron, 1979a) sidesteps the mathematical difficulties in analyzing many statistics having non-normal distributions. Bootstrap has been used to derive the sampling variance and confidence intervals of measures of niche overlap (Mueller and Altenberg, 1985), similarity index (Smith et al., 1986), Gini coefficient of inequality (Dixon et al., 1987) and growth rate (Meyer et al., 1986).
In this paper we examine the statistical accuracy (standard error, bias, mean squared error) and provide methods for estimating confidence intervals for three indices of spatial dispersion pattern.

MATERIALS AND METHODS

Artificial populations

Three types of spatial patterns of populations were generated i) random, ii) regular, and iii) contagious. The study area was represented by a unit square in each case. The random population was generated by simply drawing random numbers from a uniform distribution for the X and Y coordinates. The population consisted by 500 individuals distributed uniformly at random. The regular population was generated by locating points (individuals) on a square lattice, the exact location being stochastic and depended upon the desired uniformity. The parameter \( \sigma^2 \) created the stochasticity (cf. Hettshe and Bitz, 1979). The value of \( \sigma^2 \) was set at 0.001 and 0.002. The three contagious (clumped) populations were generated using Diggle et al.'s (1987) Modified Thomas Process. Each population consisted of \( \lambda_1 = 25 \) randomly distributed mother plant with \( \lambda_2 = 20 \) poisson distributed offsprings around each parent. The tightness of clumping depended on \( \sigma^2 \) which was set at 0.01, 0.015 and 0.02. Details of the process have appeared in Helshe and Forrester (1983). Each population was sampled by a grid of 100 quadrats (0.1 x 0.1 unit).

The Jackknife method

The jackknife estimations applied to an index of pattern detection is as follows. Let \( x_1, x_2, \ldots, x_N \) be the observed distribution of measurements (counts) in N quadrats calculate an index of pattern detection, \( P_o \). Remove ith quadrat from the pool by successively the index \( P_i \). Repeat this by successively removing one quadrat at a time from \( i = 1, \ldots, N \) quadrats. This provides N pseudovalues as follows:

\[ P_i = NP_o - (N - 1) P_i \quad i = 1, 2, \ldots, N \]

Then the jackknife estimate (\( \hat{P}_j \)) is

\[ \hat{P}_j = \frac{\sum P_i}{N} \]

and its variance is given by

\[ \text{var} (\hat{P}_j) = \sum_{i=1}^{N} \frac{(P_i - \hat{P}_j)}{N (N - 1)} \]

The approximate 100 (1 - \( \alpha \))% confidence interval is computed as follows:

\[ \hat{P}_j \pm t_{N - 1, \alpha} [\text{Var} (\hat{P}_j)]^{1/2} \]

While some authors use t distribution (Gray and Schucany, 1972; Smith et al., 1986) others proper a standard normal variate Z (Woodwald and Schucany, 1977; Adams and McCune, 1979) for establishing confidence interval.

The Bootstrap method

The bootstrap method (Efron, 1979a,b) uses the sampled observations \( x_i \), the distribution function. This probability distribution assigns mass 1/N to each \( x_i \). From the empirical distribution samples of size N are repeatedly drawn sampling with replacement. These samples are known as bootstrap samples. The test statistic, e.g., on index of pattern detection \( P \) is computed for each of the B bootstrap samples; call these \( P_{(i)} \). the number of bootstrap samples \( B \) must be large simulation. For artificial populations \( B \) was set to 100 so as to make the results of jackknife and bootstrap procedures comparable. The bootstrap estimate of \( P \) and its sampling variance are obtained as:
CONFIDENCE INTERVAL FOR BOOTSTRAP ESTIMATED WAS ESTABLISHED IN TWO WAYS A) SYMMETRIC INTERVAL AND B) ASYMMETRIC INTERVAL. THE SYMMETRIC INTERVAL IS GIVEN BY:

\[ \hat{P}_B = \frac{1}{N} \sum_{i=1}^{B} \hat{P}_{(i)} \]

AND

\[ \text{Var}(\hat{P}_B) = (B - 1)^{-1} \sum_{i=1}^{B} [\hat{P}_{(i)} - \hat{P}_B]^2 \]

THE BIAS IN ESTIMATING \( \hat{P} \) WAS FOUND AS BIAS(\( \hat{P} \)) = \( \hat{P}_B - P_0 \), THE BIAS ADJUSTED VALUE WAS FOUND AS:

\[ \hat{P}^* = \hat{P}_0 - \text{Bias}(\hat{P}) = 2\hat{P}_0 - \hat{P}_B \]

CONFIDENCE INTERVAL FOR BOOTSTRAP ESTIMATED WAS ESTABLISHED IN TWO WAYS a) SYMMETRIC INTERVAL AND b) ASYMMETRIC INTERVAL. THE SYMMETRIC INTERVAL IS GIVEN BY:

\[ \hat{P} \pm \Phi^{-1}(1 - \alpha/2) \sqrt{\text{Var}(\hat{P}_B)} \]

THE ASYMMETRIC CONFIDENCE INTERVAL BY PERCENTILE METHOD REQUIRES 1000 OR MORE REPlications. THE BOOTSTRAP SAMPLES ARE ORDERED AND THE LIMITS (L, U) OF THE 100 (1 - \( \alpha \))% CONFIDENCE INTERVAL ARE GIVEN BY 100 (\( \alpha/2 \)) AND 100 (1 - \( \alpha/2 \)) PERCENTILES OF THE ORDERED VALUES.

RESULTS

ARTIFICIAL POPULATIONS

A SUMMARY OF THE RESULTS OF JACKKNIFE ESTIMATION OF PATTERN DETECTION INDICES FOR THE ARTIFICIAL POPULATIONS ARE GIVEN IN TABLE 1. BOTH MORISITSA’S INDEX (I\( \delta \)) AND INDEX OF PATCHINESS WERE SUBSTANTIALLY GREATER THAN 1 FOR CONTAGIOUS POPULATIONS, LESS THAN 1 FOR REGULAR POPULATIONS AND DOWSE TO 1 FOR THE RANDOM POPULATION AS EXPECTED. LLOYD’S INDEX M WHICH MEASURES THE AMOUNT BY WHICH VARIANCE MEAN RATIO EXCEEDS UNITY ADDED TO MEAN DENSITY WAS HIGH FOR THE CONTAGIOUS POPULATIONS LOWER THAN THE MEAN DENSITY FOR REGULAR POPULATIONS AND CLOSE TO MEAN DENSITY FOR THE RANDOM POPULATION. ALL THREE INDICES YIELDED HIGHER VALUE FOR THE CONTAGIOUS POPULATION WITH TIGHT CLUMPING (\( \sigma^2 = 0.01 \)) AND THE VALUES DECLINED WITH THE INCREASE IN \( \sigma^2 \), I.E., WITH THE DECREASE IN CLUMPING INTENSITY.

IN GENERAL, THE JACKKNIFE ESTIMATES OF ALL THE THREE INDICES OF PATTERN DETECTION WERE CLOSE TO THE ACTUAL (POPULATION) VALUE AND THE PERCENTAGE BIAS WAS SMALL. GENERALLY THE PERCENTAGE BIAS WAS LOWEST FOR MORISITSA’S INDEX, FOLLOWED BY INDEX OF PATCHINESS AND LLOYD’S INDEX OF MEAN CROWDING IN THAT ORDER (TABLE 1). THE BOOTSTRAP ESTIMATES OF MORISITSA’S INDEX OF PATCHINESS HAD GENERALLY LOWER BIAS THAN THAT OF JACKKNIFE ESTIMATE WHILE THE REVERSE WAS TRUE FOR LLOYD’S INDEX OF MEAN CROWDING. VARIANCE WAS FOUND CLOSELY SIMILAR FOR MORISITSA’S INDEX AND INDEX OF PATCHINESS AND WAS MUCH LOWER THAN THAT OF INDEX OF MEAN CROWDING. THE VALUES OF MEAN SQUARED ERROR WERE USUALLY CLOSE TO THOSE OF THE CORRESPONDING VARIANCE DUE TO LOW BIAS.

TABLE 2 GIVES THE RESULTS OF BOOTSTRAP ESTIMATION OF PATTERN DETECTION INDICES FOR THE ARTIFICIAL POPULATIONS. PERCENTAGE BIAS WAS USUALLY LOWER FOR THE BOOTSTRAP COMPARED TO THAT OF JACKKNIFE ESTIMATE (TABLE 1 AND 2). WHERE EVER BIAS WAS LOWER THE VARIANCE OF BOOTSTRAP ESTIMATE WAS SLIGHTLY HIGHER THAN THAT OF THE JACKKNIFE. MEAN SQUARED ERRORS FOR THE PATTERN INDICES OF THE SIX ARTIFICIAL POPULATIONS WERE EITHER VERY CLOSE FOR THE TWO ESTIMATION METHODS OR SLIGHTLY LOWER FOR THE BOOTSTRAP METHOD. THE BIAS, THE VARIANCE AND THE MEAN SQUARED ERROR WERE LOWER FOR REGULAR AND RANDOM POPULATIONS COMPARED TO CONTAGIOUS POPULATION IN BOTH THE METHODS OF ESTIMATION.

BOOTSTRAP SYMMETRIC CONFIDENCE INTERVALS WERE USUALLY SLIGHTLY NARROWER THAN THOSE OF JACKKNIFE. THIS IS PARTICULARLY MORE APPARENT FOR INDEX OF PATCHINESS (TABLE 1 AND 2). BOOTSTRAP ASYMMETRIC CONFIDENCE INTERVALS WERE USUALLY FOUND TO BE CLOSE TO THE SYMMETRIC INTERVALS, IN SOME CASES THROUGH THE FORMER WERE NARROW. SIMULATIONS WERE NOT PERFORMED TO CHECK THE PERCENTAGE OF TIME THE KNOWN PARAMETER VALUE WAS CONTAINED WITHIN THE INTERVAL. HOWEVER, NORMALITY OF BOOTSTRAP SAMPLES WAS TESTED USING GEARY’S TEST OF NORMALITY (D’AGOSTINO, 1970). WITH THE EXCEPTION OF LLOYD’S INDEX OF MEAN CROWDING FOR REGULAR POPULATION WITH \( \sigma^2 = 0.002 \) (WHERE THERE WAS A SIGNIFICANT DEPARTURE FROM NORMALITY) THE BOOTSTRAP SAMPLES FOR THE ARTIFICIAL POPULATIONS DID NOT DEPART SIGNIFICANTLY FROM NORMALITY. FIG. 1 SHOWS THE HISTOGRAM OF BOOTSTRAP REPLICATIONS FOR THE THREE INDICES OF PATTERN DETECTION IN RELATION TO THE CONTAGIOUS POPULATION WITH \( \sigma^2 = 0.01 \). ALL THE THREE INDICES SHOWS NORMAL DISTRIBUTION.
Table 1. Results of jackknife estimation of three pattern detection indices for the artificial populations. Given below are percent bias, variance, mean squared error (MSE) and 95% confidence interval (C.I.).

<table>
<thead>
<tr>
<th>Populations</th>
<th>Index</th>
<th>Calculated by index</th>
<th>Jackknife estimate</th>
<th>% Bias</th>
<th>Var (P3)</th>
<th>MSE</th>
<th>95% C.I.</th>
</tr>
</thead>
<tbody>
<tr>
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<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Contagious δ=0.01</td>
<td>I₀</td>
<td>1.9464</td>
<td>1.9629</td>
<td>0.8486</td>
<td>0.03210</td>
<td>0.03211</td>
<td>1.6074</td>
</tr>
<tr>
<td></td>
<td>m²</td>
<td>9.8099</td>
<td>9.9171</td>
<td>1.09327</td>
<td>1.78521</td>
<td>1.78532</td>
<td>7.2662</td>
</tr>
<tr>
<td></td>
<td>C</td>
<td>1.9425</td>
<td>1.9629</td>
<td>1.04944</td>
<td>0.03196</td>
<td>0.03197</td>
<td>16085</td>
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<tr>
<td>Contagious δ=0.015</td>
<td>I₀</td>
<td>1.7062</td>
<td>1.7091</td>
<td>0.1735</td>
<td>0.01451</td>
<td>0.01454</td>
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<td></td>
<td>m²</td>
<td>8.7187</td>
<td>8.7743</td>
<td>0.6373</td>
<td>0.65449</td>
<td>0.65452</td>
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<tr>
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<td>1.7028</td>
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<td>0.315</td>
<td>0.01442</td>
<td>0.01442</td>
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<td>I₀</td>
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<td>1.4847</td>
<td>0.19488</td>
<td>0.00775</td>
<td>0.00775</td>
<td>1.3100</td>
</tr>
<tr>
<td></td>
<td>m²</td>
<td>7.6763</td>
<td>7.7171</td>
<td>0.5322</td>
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<td>0.43404</td>
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<td></td>
<td>C</td>
<td>1.4791</td>
<td>1.4848</td>
<td>0.3895</td>
<td>0.00770</td>
<td>0.00770</td>
<td>1.3107</td>
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<tr>
<td>Regular δ=0.001</td>
<td>I₀</td>
<td>0.8986</td>
<td>0.8977</td>
<td>0.1039</td>
<td>0.000153</td>
<td>0.000153</td>
<td>0.8731</td>
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<td>m²</td>
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<td>4.7488</td>
<td>0.0861</td>
<td>0.0235</td>
<td>0.0235</td>
<td>4.4446</td>
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<td>0.8977</td>
<td>0.0855</td>
<td>0.000152</td>
<td>0.000152</td>
<td>0.8732</td>
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<td>0.9194</td>
<td>0.9188</td>
<td>0.0615</td>
<td>0.000195244</td>
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<td>C</td>
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<td>0.9188</td>
<td>0.1280</td>
<td>0.0019483</td>
<td>0.0019483</td>
<td>0.8911</td>
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Table 2. Results of bootstrap estimation (B = 1000) of three pattern detection indices for the artificial populations. Given below are percent bias, variance, mean squared error (MSE) and 95% confidence interval (C.I.).

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<tr>
<th>Populations</th>
<th>Index</th>
<th>Calculated by index</th>
<th>Bootstrap estimate</th>
<th>% Bias</th>
<th>Var (P3)</th>
<th>MSE</th>
<th>95% C.I.</th>
</tr>
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<tbody>
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<td></td>
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<tr>
<td>Contagious δ=0.01</td>
<td>I₀</td>
<td>1.9464</td>
<td>1.9339</td>
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<td>9.8099</td>
<td>9.6963</td>
<td>1.1577</td>
<td>1.60306</td>
<td>1.61596</td>
<td>7.3283</td>
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<td>C</td>
<td>1.9425</td>
<td>1.9300</td>
<td>0.6447</td>
<td>0.03387</td>
<td>0.03403</td>
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<td>1.7037</td>
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<td>0.01437</td>
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<td>8.7185</td>
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<td>0.0084</td>
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<td>0.00014</td>
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<td>Regular δ=0.002</td>
<td>I₀</td>
<td>0.9194</td>
<td>0.9192</td>
<td>0.0141</td>
<td>0.00022</td>
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</table>

DISCUSSION

For many indices used in ecology such as diversity, similarity, niche overlap, remains unknown. Computer simulation in such situations is one alternative (e.g., Ricklefs and Lau, 1980; Smith and Zaret, 1982). This is in
essence bootstrap estimation of the parameters. Both bootstrap and jackknife provide the alternative means for determining the statistical accuracy.

The results showed that both the jackknife and bootstrap are effective at reducing bias and at lower sample size bootstrap is more effective in this respect. It is evident, however, that there is a cost to reducing bias in the form of increased variance of the estimator. The bias, variance and mean squared errors are all reduced when the populations are either regular or random. Thus, the statistical accuracy of indices of pattern detection are greater when the populations are distributed are greater when the populations are distribution regularly or randomly and lower when the populations follow aggregated distributions.

Even though the symmetric jackknife and bootstrap confidence interval estimates were calculated differently the results were usually closely similar. For constructing the jackknife interval the conservative to t-variante.

Schenker (1985) has pointed out that bootstrap confidence intervals should be used with caution in complex problems was used while for bootstrap interval Z-variate was employed. The accuracy of confidence intervals was not checked by simulation. However, these interval estimates are expected to be accurate when the resampled values (jackknife pseudovalues and bootstrap samples) are normally distributed. Geary’s test of normality showed that with a few exceptions they in fact followed normal distributions for the populations considered. The asymmetric confidence interval based on percentile method also yielded closely similar results to those of jackknife and bootstrap. In the context of a similarity index Smith et al., (1986) demonstrated that the bootstrap percentile method works well regardless of the skewness. As the number of observations increases, the frequency distribution of bootstrap value approaches the frequency distribution of original data (Nash, 1981). Therefore, bootstrap interval estimates are expected to be more accurate for the larger data sets. Besides the bootstrap confidence interval methods used in this study other methods have also been proposed (cf. Efron and Tibshirani, 1986; Hall, 1988; DiCiccio and Romano, 1990). On the basis of our results based on natural and artificial populations it appears that both jackknife and bootstrap perform more or less equally well when the sample size is small while bootstrap out performs jackknife when the sample size is large, particularly for Moristia’s index and index of patchiness.

In ecology as well as in certain other disciplines it is often required to compare the dispersion pattern of two or more populations. Once the variances of an index of pattern detection of two populations are known by either jackknife or bootstrap method the significance of difference in their dispersion pattern can easily be tested using the familiar Z-statistic.

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