

CONCLUSIONS

Analysis of count data should begin with the frank acknowledgment that counts are not necessarily very good surrogates for population sizes. Pattern in counts

scripted (the selected model had a sixth-degree polynomial log-trajectory and an intervention effect corresponding to the severe winter of 1976-1977, for a seven-parameter model). Fifteen of the 20 BBS routes in the Maryland Northern Piedmont stratum had adequate data to fit the seven-parameter model. We fit the

initial modeling, biological intuition, and familiarity with the methods of the surveys producing the counts. For the BBS, we have shown that several factors produce pattern in the proportion of birds counted; these include differences among observers and changes through time in the ability of individual observers (e.g., Kendall, Link, and Sauer 1996) to accommodate Link and Sauer's (1996) survey design. These factors and other sources of variation are likely to be important in appropriate modeling.

Link and Sauer's (1996) approach to modeling individual routes or regions is useful for the collection, acknowledging the differences in precision of the estimates (Link and Sauer 1996). Their empirical Bayes estimates can then be averaged with weights of relative abundance and used to provide an estimated trend for the population total.

Spatial comparisons of relative abundance from BBS data must be viewed with some skepticism. The fundamental difference between spatial and temporal analysis of BBS data is that the assumption of consistency in detectability can be plausibly advanced in considering counts for individual observers taken at the same site, through time; this assumption is less plausible for comparisons of counts among routes at a large geographic scale. Regional differences in detectability of birds may exist, with the potential to bias estimates of relative abundance (Sauer et al. 1993). Unfortunately, little information is available on counts for individual observers at distant sites; hence, this component of detectability has never been modeled.

CONCLUSIONS

Analysis of count data should begin with the frank acknowledgement that counts are not necessarily very good surrogates for population sizes. Pattern in counts is indicative and only of corresponding patterns in population sizes, but also of corresponding patterns in the proportion of animals counted. Survey producing count data should be designed, instead of as possible, to minimize variation in the proportion of animals counted. Complete removal of this variation is not likely to be possible, hence, analysis must be aware of potential sources of pattern in this proportion, and must design analytic methods accordingly. Thus, analysis of count data requires a delicate interaction among sta-

tistical modeling, biological intuition, and familiarity with the methods of the surveys producing the counts. For the BBS, we have shown that several factors produce pattern in the proportion of birds counted; these include differences among observers and changes through time in the ability of individual observers (e.g., Kendall, Link, and Sauer 1996) to accommodate Link and Sauer's (1996) survey design. These factors and other sources of variation are likely to be important in appropriate modeling.

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LITERATURE CITED

Baker, R. J., and J. R. Sauer. 1992. Modeling population change from time series data. Pages 182-194 in D. R. McCullough and R. Barrett, editors. *Wildlife 2001: population projections*. New York, USA.

Flether, C. H., and J. R. Sauer. 1996. Using landscape ecology to test hypotheses about large-scale abundance patterns in migratory songbirds. *Ecology* 77:28-35.

Gensler, P. H., and B. R. Scott. 1981. Estimates of avian population trends from the North American Breeding Bird Survey. Pages 42-51 in C. J. Ralph and J. M. Scott, editors. *Estimating numbers of terrestrial birds: Studies in Avian Biology*, 6.

Gensler, P. H., and J. R. Sauer. 1990. Topics in comparative avian analysis. Pages 54-57 in J. R. Sauer and S. Dove, editors. *Survey designs and statistical methods for the estimation of avian population trends*. U.S. Fish and Wildlife Service, Biological Report 90(1).

Harris, T. J., and R. J. Thorburn. 1990. Geographical additive models. Chapman and Hall, New York, USA.

Jones, F. C., C. E. McCulloch, and D. A. Wiedenfeld. 1996. New approaches to the analysis of Population trends in land birds. *Ecology* 77:13-21.

Kendall, W. L., G. Petroulou, and J. R. Sauer. 1996. First-time observer effects in the North American Breeding Bird Survey. *Auk* 113:433-439.

Lauber, R. A., J. D. Nichols, and K. H. Pollock. 1994. Estimating the number of animals in wildlife populations.