

CTA vs. Chi-Square: Comparing Voter Sentiment in Political Wards

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Random samples of 200 registered voters from each of four political wards were asked if they favored a particular candidate.¹ Seven chi-square analyses (one omnibus comparison between all four wards, six follow-up pair-wise comparisons to specify the underlying effect²) were used to compare the proportion of voters favoring the candidate between wards. Evaluating results at either the generalized or the experimentwise criterion for statistical significance, chi-square found the omnibus effect ($p<0.013$), and two pairwise comparisons: Ward 1>Ward 2 ($p<0.0025$), and Ward 1>Ward 4 ($p<0.014$). In contrast, a single CTA analysis was conducted predicting voter sentiment (treated as the class variable, and coded as 1 if the voter favors the candidate, or 0 otherwise) with ward (dummy-coded as 1-4) treated as a multcategorical attribute.³⁻⁹ A single model emerged: if Ward=1, then predict the voter favors the candidate; otherwise predict the voter does not favor the candidate ($p<0.036$). The training (total sample) effect was relatively weak (ESS=10.2), and the predictive accuracy declined to levels worse than expected by chance (ESS= -14.8) in jackknife analysis. CTA thus revealed that the most accurate model possible for this application is weak, and that there is evidence that the model may not cross-generalize if it is used to classify independent random samples.

Data investigated herein are given in Table 1.

Table 1: Study Data¹

Favor	Yes	Ward 1	Ward 2	Ward 3	Ward 4
Candidate	No	76	53	59	48
		124	147	141	152

In training (total sample) analysis the CTA model correctly classified 440/564 (78.0%) of the voters opposing the candidate, and 76/236

(32.2%) of the voters supporting the candidate. However, in jackknife analysis model sensitivity fell to 299/564 (53.0%) for voters who were against the candidate. ESS cannot ordinarily be computed for findings for multcategorical designs that are based on chi-square, and can never be computed unless the findings are completely specified. For example, here chi-square finds Ward 1>Ward 2 and Ward 1>Ward 3: how do Ward 2 and Ward 3 compare, and what about Ward 4?

References

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Author Notes

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