Blue and Red Counties in the 2008 Presidential Election: An Analysis of Intracluster Correlation

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Abstract
Recent theories suggest that individuals are clustering within culturally and politically homogeneous neighborhoods. This clustering has led to an increasingly polarized electorate comprised of individuals who have an increased intolerance for ideas and behaviors that are different from their own (Bishop, 2008).

The 2008 American National Election Time Series Study (ANETSS), a nationally-representative political behavior and public opinion survey, is based on a probability sample of 82 Counties. Using intracluster correlation as a measure of “sameness” on key outcomes (e.g., political ideology, voting behavior, religious affiliation, and political enthusiasm), we evaluate whether or not the 2008 ANETSS data support the clustering theory within counties.

Key Words: Political Polarization, Cluster Analysis, 2008 American National Election Time Series Study

1. A Big Sort?

Recent theories suggest that individuals have been clustering themselves into culturally and politically homogeneous places (Bishop, 2008). Over the past several decades, people across the country have migrated into culturally-distinct neighborhoods, resulting in a political segregation. Communities across the United States have become more homogeneous in cultural values, religious beliefs, and political attitudes. In The Big Sort, Bill Bishop claims that this clustering has led to an increasingly polarized electorate comprised of individuals who have an increased intolerance for ideas and behaviors that are different from their own.

1.1 Level of Clustering

Under the electoral system in the United States, the presidential candidate receiving the plurality of each state’s votes wins all of the electoral votes for that state. This ‘winner takes all’ system means that much attention is given to the concept of red states (voting predominately for Republican candidates) and blue states (voting predominately for Democratic candidates) (see Exhibit 1). Looking at county vote returns within each state, it is evident that political segregation is occurring not at the state level, but at the county or sub-county level (see Exhibit 2). If we define ‘landslide’ blue counties as those where more than 60% of voters voted for Barack Obama in the 2008 Presidential election and ‘landslide red’ counties as those where more than 60% of voters voted for John McCain, then we see that there are landslide blue counties in red states and landslide red counties in blue states (see Exhibit 3). This clustering likely extends to the sub-county level, where

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1 Nebraska and Maine do not follow the ‘winner takes all’ rule.
blue neighborhoods exist within red counties and red neighborhoods exist within blue counties.

**Exhibit 1:** 2008 Presidential Vote Results by State

![Map of 2008 Presidential Vote Results by State](http://elections.nytimes.com/2008/results/president/map.html)

<table>
<thead>
<tr>
<th>State Color</th>
<th>County Outcome</th>
<th>Number of Counties</th>
<th>Percent</th>
</tr>
</thead>
<tbody>
<tr>
<td>Blue</td>
<td>Landslide Blue</td>
<td>200</td>
<td>6%</td>
</tr>
<tr>
<td></td>
<td>Landslide Red</td>
<td>251</td>
<td>8%</td>
</tr>
<tr>
<td></td>
<td>Other</td>
<td>969</td>
<td>31%</td>
</tr>
<tr>
<td></td>
<td>Total</td>
<td>1,420</td>
<td>46%</td>
</tr>
<tr>
<td>Red</td>
<td>Landslide Blue</td>
<td>99</td>
<td>3%</td>
</tr>
<tr>
<td></td>
<td>Landslide Red</td>
<td>1,072</td>
<td>34%</td>
</tr>
<tr>
<td></td>
<td>Other</td>
<td>523</td>
<td>17%</td>
</tr>
<tr>
<td></td>
<td>Total</td>
<td>1,694</td>
<td>54%</td>
</tr>
</tbody>
</table>

**Exhibit 2:** 2008 Presidential Vote Results by County

**Exhibit 3:** Distribution of Landslide Counties in the 2008 Presidential Election

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1.2 Characteristics that Define Clustering

Several demographic characteristics define the clustering of individuals into politically and culturally homogeneous communities (Bishop 2008). In recent elections, population density has been a strong predictor of political preference. Republicans have been migrating to rural areas of the country while Democrats have been migrating to more urban areas.

Education, income, and age have also become defining characteristics of the clustering. People with post-secondary education have been migrating to cities with high concentrations of people with higher education degrees, creating both income disparities and disparities in educational attainment among different cities across the country. This trend is even more pronounced for young Americans, and the disparities are particularly striking when looking at urban versus rural areas. Young adults in rural areas are less likely to have a college degree than young adults in urban areas.

There is also a religious component to the clustering. Democratic counties have been losing churchgoers in recent years, while church membership has been increasing in predominately Republican counties. This has political implications. Those who attended church at least once a week were more likely to vote Republican in 2000 and 2004.

Additionally, there are racial and marital components to the clustering. Since 1970, white Americans have become increasingly more concentrated in counties voting predominately for Republicans. A higher percentage of people are married in predominately Republican counties than in predominately Democratic counties.

2. The 2008 American National Election Time Series Study

2.1 Survey Design

The 2008 American National Election Time Series Study (ANETSS) is a nationally-representative survey of political behaviour and public opinion. It was based on a multi-stage probability sample consisting of 82 counties, 160 Census Tracts, 297 Census Block Groups (CBGs), and resulted in 2,323 pre-election interviews. The 2008 ANETSS oversampled Latinos and African Americans.

The ANETSS data were weighted to reflect the different selection probabilities at various stages of sampling and to compensate for differential nonresponse and undercoverage. The weighting process entailed three major steps. The first step consisted of the computation of design weights to account for unequal probabilities of selection at each stage. In the second step, the design weights were adjusted for nonresponse using a response propensity approach (Folsom 1991). In the third step, the nonresponse-adjusted weights were poststratified to Current Population Survey estimates of the target population of U.S. citizens 18 years of age and older to ensure proper coverage.

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4 Landslide Counties are those where over 60% of the voters voted for either Barack Obama (Landslide Blue Counties) or John McCain (Landslide Red Counties).


We used the *PROC WTADJUST* procedure in SUDAAN (RTI 2008) to adjust the design weights for nonresponse, undercoverage, and to truncate extreme weights. The procedure implements the Generalized Exponential Model of Folsom and Singh (2000) which provides double protection against the biases from nonresponse and coverage error because its use can be justified with either a coverage model or with a response prediction model. Extreme weights were trimmed in a way that any losses/gains in the weight sums were accounted for in the subsequent computation of the weight adjustments.

2.2 The 2008 ANETSS and Political Clustering

Since the primary sampling units (PSUs) in the 2008 ANETSS were counties and since the ANETSS measures political and cultural outcomes related to the factors contributing to the political clustering described in *The Big Sort*, both the design and the content of the ANETSS enabled a two-part analysis of the political clustering theory. First, we examined relevant study outcomes to determine whether or not political clustering existed within counties. Then, we attempted to explain within-county homogeneity using demographic factors such as education, age, and race.

We first assigned a color to each county using auxiliary data. We assigned each of the 82 counties in the sample a county color based on the actual vote returns for that county (see Exhibit 4). These color assignments consisted of landslide victory counties (very blue and very red), solid victory counties (blue and red), and battleground counties (purple).

<table>
<thead>
<tr>
<th>County Color</th>
<th>County Vote Results</th>
</tr>
</thead>
<tbody>
<tr>
<td>Very Blue</td>
<td>% Obama &gt; 60%</td>
</tr>
<tr>
<td>Blue</td>
<td>52% ≤ % Obama ≤ 60%</td>
</tr>
<tr>
<td>Purple</td>
<td>% Obama &lt; 52% and % McCain &lt; 52%</td>
</tr>
<tr>
<td>Red</td>
<td>52% ≤ %McCain ≤ 60%</td>
</tr>
<tr>
<td>Very Red</td>
<td>% McCain &gt; 60%</td>
</tr>
</tbody>
</table>

*Exhibit 4:* Assignment of County Color

**Did the ANETSS Track the 2008 Election Results?**

Before using the results of the 2008 ANETSS to determine whether or not political and cultural clustering exists in the United States, we first wanted to ensure that the 2008 ANETSS adequately represented the population. Despite the over-sample of Latinos and African-Americans, weighted estimates from the 2008 ANETSS tracked the overall election results very closely (see Exhibit 5).

<table>
<thead>
<tr>
<th>Candidate</th>
<th>Estimate</th>
<th>Standard Error</th>
<th>Actual Popular Vote6</th>
</tr>
</thead>
<tbody>
<tr>
<td>Barack Obama</td>
<td>53.0%</td>
<td>2.9%</td>
<td>52.7%</td>
</tr>
<tr>
<td>John McCain</td>
<td>43.5%</td>
<td>2.9%</td>
<td>46.0%</td>
</tr>
<tr>
<td>Other</td>
<td>1.9%</td>
<td>0.5%</td>
<td>1.3%</td>
</tr>
<tr>
<td>Don't Know / Refused</td>
<td>1.7%</td>
<td>0.3%</td>
<td>N/A</td>
</tr>
</tbody>
</table>

*Exhibit 5:* Estimated 2008 Presidential Vote and Actual Popular Vote

6 CNN Election Center (2008)
Did the ANETSS Track the 2008 Election Results by County Color?

Since our analysis is based on a comparison of different types of counties from the 2008 ANETSS (i.e. different county colors), we wanted to ensure that the 82 counties in the sample were representative of the population in terms of the distribution of county colors. As Exhibit 6 shows, the estimated distribution of the eligible voting population from the ANETSS (Sample Distribution) tracks fairly closely to the national distribution of the voting population within each county color. For example, 27% of the voting population live in counties that voted in a landslide for Barack Obama (very blue counties), while the weighted estimate from the 2008 ANETSS is 22% with a standard error of 5%.

<table>
<thead>
<tr>
<th>County Color</th>
<th>National Distribution</th>
<th>Sample Distribution</th>
<th>Std. Error</th>
</tr>
</thead>
<tbody>
<tr>
<td>Very Blue</td>
<td>27%</td>
<td>22%</td>
<td>5%</td>
</tr>
<tr>
<td>Blue</td>
<td>25%</td>
<td>18%</td>
<td>5%</td>
</tr>
<tr>
<td>Purple</td>
<td>15%</td>
<td>18%</td>
<td>5%</td>
</tr>
<tr>
<td>Red</td>
<td>16%</td>
<td>17%</td>
<td>5%</td>
</tr>
<tr>
<td>Very Red</td>
<td>17%</td>
<td>25%</td>
<td>6%</td>
</tr>
</tbody>
</table>

Exhibit 6: Comparison of National and Sample County Color Distribution

3. A Measure of Homogeneity

Since the goal of this analysis was to compare the degree of homogeneity among different types of counties (e.g. blue counties and red counties), we needed a measure for this homogeneity.

Intracluster correlation (ICC) is a measure of homogeneity within clusters. ICC tells us how similar elements within the same cluster are on a particular outcome (Lohr 1999).

\[
ICC = 1 - \frac{M}{M-1} \cdot \frac{SSW}{SSTO}, \text{ where}
\]

\[
M = \text{cluster size}
\]
\[
SSW = \text{Sum of Squares within PSUs}
\]
\[
SSTO = \text{Total Sum of Squares}
\]

Thus a high ICC indicates a high level of homogeneity within clusters since most of the total variance is due to variation between clusters rather than variation within clusters.

ICC is a better measure of homogeneity than simply comparing means and confidence intervals since ICC tells us the source of the variation (between clusters versus within clusters). Means and standard errors can be similar with very different ICCs. This is evident in Exhibit 7. The estimated proportions of people in very blue counties who felt

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7 The national distribution excludes Alaska and Hawaii since they were excluded from the target population of the 2008 ANETSS
8 All ICC estimates were made using SUDAAN’s (RTI 2008) PROC LOGISTIC and PROC REGRESS
angry or afraid of Barack Obama were similar, but we estimated more within-county agreement on the ‘afraid’ question than the ‘angry’ question.

<table>
<thead>
<tr>
<th>Outcome</th>
<th>Mean</th>
<th>Std. Error</th>
<th>ICC</th>
</tr>
</thead>
<tbody>
<tr>
<td>Angry</td>
<td>20%</td>
<td>2%</td>
<td>0.04</td>
</tr>
<tr>
<td>Afraid</td>
<td>19%</td>
<td>3%</td>
<td>0.10</td>
</tr>
</tbody>
</table>

Exhibit 7: Comparison of Means, Standard Errors, and ICCs for Two Outcomes from the 2008 ANETSS in Very Blue Counties

4. Analysis and Results

We calculated the ICCs within each of the five county colors for 26 political attitude, political behavior, and cultural behavior outcomes from the 2008 ANETSS. We expected that ICC would be the highest in landslide counties (very blue and very red), moderate in solid victory counties (blue and red), and lowest in battleground counties (purple). This would produce a U-shaped graph as we moved from very blue to very red counties.

4.1 Single Outcome ICCs

Many outcomes from the 2008 ANETSS demonstrated the U-shaped ICCs. Exhibit 8 shows the support for Barack Obama (the respondent either voted for Obama or, if the respondent did not vote, the respondent intended to vote for Obama). As would be expected, the proportion of support for Obama decreased from very blue to very red counties (from 73% to 37%). However, the ICCs follow a U-shaped pattern, with the strongest levels of homogeneity in the very blue and very red counties, and weakest homogeneity demonstrated in the purple counties.

Exhibit 8: Support for Barack Obama: Means and ICCs in the 2008 ANETSS

While many outcomes that we expected to demonstrate within-county homogeneity did so, other outcomes did not demonstrate the homogeneity that we expected. For example, weekly church attendance across the different county colors did not demonstrate
noticeable differences, either in the percent who attend church once a week or on the within-county homogeneity (ICCs) on this issue (see Exhibit 9).

Exhibit 9: Church Attendance: Means and ICCs in the 2008 ANETSS

Averaged across all 26 individual outcomes of interest, the ICCs within each county color demonstrate the expected U-shape (see Exhibit 10). While it is clear from the minimum ICC within each county color that some outcomes demonstrated little to no clustering, as a whole the 2008 ANETSS data did support the clustering theory. However, ICCs in general were lower than expected. The maximum ICC across the 26 outcomes for any county color was 0.26, which means that an estimated 74% of the variation on this outcome was attributable to within-county variation. This within-county heterogeneity might be explained by sub-county clustering (i.e. there are red neighborhoods in blue counties and blue neighborhoods in red counties).

Exhibit 10: ICCs for 26 Outcomes of Interest from the 2008 ANETSS

4.2 Explaining the Homogeneity

In The Big Sort, Bill Bishop proposes several outcomes that contribute to the clustering of individuals within like-minded communities. To test whether or not these variables ‘explain away’ the homogeneity, we fit both the unadjusted model demonstrated in section 4.1 (Obama = county color) and an adjusted model using variables related to the covariates Bishop cited as defining factors of the homogeneity (Obama = County Color, Race, Urbanicity, Education, Marital Status, Age, Church Attendance). The results demonstrate that the covariates account for almost all of the homogeneity within blue,
purple, and red counties. However, there is still some homogeneity left unexplained in the very blue and very red counties (see Exhibit 11).

Exhibit 11: Unadjusted ICC vs. ICC Adjusted for County Color, Race, Urbanicity, Education, Marital Status, Age, and Church Attendance

5. Conclusion and Future Work

For some attitudes and behaviors, such as feelings towards Barack Obama and John McCain, the data in the 2008 ANETSS show evidence of political clustering. The data show higher levels of within-county homogeneity in landslide counties (counties that are more politically clustered) than competitive counties (counties that are not as politically clustered) for several outcomes of interest. However, the data show that some attitudes that we would expect to demonstrate within-county homogeneity are not homogeneous within counties, such as church attendance and feelings about same-sex marriage. Other attitudes are more homogeneous in red counties than blue counties (e.g. feelings about whether or not the country is headed in the right direction) or in blue counties than red counties (e.g. feelings towards Christian Fundamentalists).

Education, race, urbanicity, age, and marital status explain much of the within-county homogeneity for support for Obama, but the remaining homogeneity in the landslide counties and the relatively low ICCs in general demonstrate the need to look below the county level.

Future work will focus on neighborhood homogeneity. Census Block Groups (CBGs) are geographic areas that, on average, contain about 500 households, so CBGs can be thought of as neighborhoods. The 2008 ANETSS data come from 297 CBGs, which allows for analysis at the neighborhood level. We will utilize Hierarchical Linear Modeling (HLM) as a way to estimate the variance components attributable to counties and to CBGs within
counties. We will also use a Jackknife method to calculate variances for the ICCs to utilize formal testing between the county colors on various outcomes of interest.

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References


