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Abstract	<p>Recent publications have recognized bias in voter and party placements such as “differential item functioning” (in a broader sense) or “projection effects” (in a narrower sense). This paper is concerned with the latter, and we suggest both a measurement model and a re-scaling approach that corrects for projection bias in voter-specific party placements. Assessments of spatial voting models are systematically distorted when voters tend to pull parties they prefer closer to their ideal points (“assimilation”) while pushing parties they dislike farther away (“contrast”). These projection effects tend to contaminate voter perceptions of spatial positions with non-spatial considerations and thus introduce bias to empirical evaluations of alternative models of vote choice.</p> <p>The paper proceeds in two consecutive steps: In the first part, we utilize a large, cumulative dataset compiled from the “Comparative Study of Electoral Systems” (CSES) survey series to demonstrate and measure the presence, magnitude, and contextual variation of projection effects, i.e., of assimilation and contrast. Building on these empirical findings, the second part suggests a re-scaling technique that corrects for systematically biased party placements. We apply these techniques to a subset of the cumulative CSES data to highlight the practical added value of obtaining rescaled and unbiased party positions.</p>	
Keywords (separated by '-')	Spatial model - Political psychology - Vote choice - Projection effects - Scaling techniques	



Projection Effects: Coping with Assimilation and Contrast

Guido Tiemann

1 Introduction

The spatial model of voting is often considered *the* workhorse of modern electoral studies and for decades belongs to the standard toolbox of empirical election studies (cf. Downs 1957; Davis et al. 1970; Enelow and Hinich 1984). In principle, any spatial voting model claims two things: (i) there is a meaningful relationship between the ideal points of voters and the programmatic positions of political parties in some one- or multidimensional political space, and (ii) these relations matter for (models of) party evaluation and vote choice.

Therefore, the validity of empirical research depends on voters' ability to meaningfully locate parties within a political space. Notable contributions have shown that there are limits to this and raised some more general concerns about "differential item functioning". Provided that issue scales in survey questionnaires are usually merely labeled at their end or extreme points, respondents may understand scales and scale points in very different ways. Even preserving the general ordering of candidates or parties along a dimension, some voters may be willing to use the entire scale, but the placements of others may be more centrist and confined to moderate positions. Empirical research has also demonstrated that respondents tend to place themselves, candidates, or parties they like at rather centrist positions. In contrast, voters tend to locate parties they cogently oppose much farther out.

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20 The electoral studies literature has repeatedly addressed these issues: notably,
21 a recent strand of literature has recurred to a scaling procedure suggested by
22 Aldrich and McKelvey (1977) to provide corrected, rescaled party placements
23 (Hare et al. 2014; Lo et al. 2014). Other contributions have focused on the
24 more specific concern that idiosyncratic party placements may be biased due
25 to alleged projection effects. Individuals may locate parties they like, for what-
26 ever reason, closer to their personal ideal point (“assimilation”), while they push
27 parties they oppose, for whatever reason, farther away (“contrast”). Subsequent
28 empirical analyses have compiled valid and robust evidence for the presence of
29 projection effects (Granberg et al. 1988; Grand and Tiemann 2013; Judd et al.
30 1983; Krosnick 1990; Miller et al. 1986; Rahn et al. 1994).

31 Whether concerns about bias and distortion in unscaled idiosyncratic party
32 placements are justified crucially depends upon the angle from which spatial
33 party positions are conceptualized and evaluated. From a subjectivist perspec-
34 tive, there is no need to correct or rescale idiosyncratic placements of candidates
35 and/ or parties. Voters are only able to apply information they are personally
36 aware of, parties are located where voters think they stand, and unified, objective
37 party placements are either fictitious or unobservable. From an objectivist per-
38 spective, idiosyncratic party placements are systematically distorted perceptions
39 of some latent, objective party location. Biased party placements therefore need
40 to be addressed to ensure correct, unbiased inferences about electoral choice.
41 Candidates and parties may (at least) be attributed some latent, true positions,
42 and voters may either correctly or incorrectly perceive these locations, and they
43 may do so with unsystematic error or with systematic bias.

44 This paper explores the specific implications of projection effects, evalu-
45 ates their presence, measures their magnitude, and studies the implications of
46 assimilation and contrast for empirical and statistical models of electoral choice.
47 Systematically distorted perceptions of party positions are not only a problem
48 per se. Biased party placements may also modify electoral behavior and lead to
49 distorted and wrong inferences by electoral researchers. Empirical evidence for
50 projection effects thus calls into question the key elements of the spatial voting
51 framework.

52 This paper is particularly concerned with three inter-related aspects: i) estab-
53 lishing the presence and measuring the magnitude of assimilation and contrast
54 across diverse, heterogeneous contexts, ii) providing a procedure for deriving
55 unbiased party placements, iii) establishing the consequences of systematic pro-
56 jection bias for the estimation of electoral choice models. We address these
57 issues in three consecutive, inter-connected steps: i) We begin with the key argu-
58 ments, develop a formal definition of assimilation and contrast, and propose a

59 measurement model for the differences between subjective and objective party
60 placements. The measurement strategy builds upon a Bayesian hierarchical model
61 and provides descriptive evidence of the presence and the magnitude of projection
62 bias across diverse contexts. ii) Because our empirical analysis provides robust
63 evidence on systematic projection bias, we build upon the previous inferences
64 and suggest a technique to correct and rescale idiosyncratic party placements.
65 iii) That said, we cannot simply assume that the presence of projection bias also
66 carries over to estimation bias in models of vote choice, i.e., that assimilation
67 and contrast make us systematically overestimate the salience of spatial proxim-
68 ity terms and underestimate the salience of alternative specifications of spatial
69 models or non-spatial utility components. We therefore compare the impact of
70 subjective, idiosyncratic party placements, objective mean party positions, and
71 re-scaled, corrected party placements in empirical models of vote choice.

72 2 Defining Projection Effects

73 Explanations for projection effects build on consistency perspectives which posit
74 that individuals try to overcome and work against cognitive dissonance. Predom-
75 inant concepts in psychology are dissonance theory (Festinger 1957), balance
76 theory (Heider 1958), and congruity theory (Osgood and Tannenbaum 1955).
77 Individuals experience cognitive inconsistency if two cognitions do not cor-
78 respond, and the perceived discomfort increases with the level of substantive
79 importance attributed to the underlying issues. The dualism of assimilation and
80 contrast is an established building block in the social psychology and cognition
81 theory research traditions.

82 These classical approaches have been adopted by electoral studies and uti-
83 lized to study voter perceptions of party ideologies and positions. Applied to
84 individual-specific party placements, voters may achieve psychological consis-
85 tency by either shifting their personal ideal points or by manipulating the political
86 parties' spatial positions. To clarify some confusion in the imminent literature,
87 we start by applying key notions from cognition theory to electoral studies:

- 88 1. The voter may adjust or change her individual spatial position due to efforts
89 taken by political parties to persuade potential voters of specific party policies
90 (“persuasion”).
- 91 2. The voter may review and modify the spatial positions taken by alternative
92 parties to bring consistency back in. From the angle of Downsian proximity
93 voting,



- 94 a) she might pull positions of political parties she likes for non-spatial reasons
95 closer to her unchanged spatial ideal point (“assimilation”), or
96 b) she might push parties she opposes for non-spatial reasons farther away
97 from her unchanged ideal point (“contrast”).

98 The inherent causal mechanisms are established with reference to classical con-
99 tributions to political psychology. Firstly, following Heider’s balance theory, the
100 individual need for establishing cognitive consistency depends on the existence
101 of “unit relations” with the respective candidate or party, i.e., the stronger the
102 attachment to a specific candidate or party, the higher the individual pressure to
103 eradicate cognitive dissonance. Secondly, dissonance theory also stresses the role
104 of choices in strengthening the need for projection effects. For instance, casting a
105 vote for a specific candidate or party increases the chance of assimilation effects
106 to eliminate any possible post-decisional cognitive dissonance. Thirdly, not only
107 electoral preferences and choices but the importance attributed to specific issues
108 determine how sensitive an individual is to cognitive inconsistencies. The salience
109 of an issue dimension therefore governs the presence and magnitude of projection
110 effects.

111 This contribution focuses on the extent and determinants of projection effects
112 rather than the electoral consequences of party cues and persuasion effects.
113 Notably, assimilation and contrast label and refer to causal effects that cannot
114 be directly observed (cf. Krosnick 1990, 2002). Conceptually, political scientists
115 are thus often not in a position to unambiguously establish whether a voter has
116 modified her individual position due to successful persuasion or whether she
117 has altered the spatial positions she assigns to the party alternatives as a result
118 of assimilation and/or contrast (cf. Conover and Feldman 1982; Tomz and van
119 Houweling 2008).

120 Krosnick (2002) argues that cross-sectional data alone is per se insufficient for
121 any clear differentiation between persuasion and projection effects. While cross-
122 sectional data analyses may often corroborate the existence of projection effects,
123 the empirical evidence is also consistent with a whole gamut of alternative expla-
124 nations, e.g. “(1) perspective effects, (2) policy-based evaluation and persuasion,
125 (3) variation in candidate’s issue statements, and (4) the agreement principle”
126 (Krosnick 2002, S. 121–122). Instead, dynamic panel models potentially facili-
127 tate the isolation of causal factors due to the temporal sequencing of events. Still,
128 panel data is often not available and comes with problems of its own.

129 Given that projection effects, assimilation and contrast, imply that non-spatial
130 motives systematically bias voter-specific party placements, our measurement
131 strategy proceeds in two consecutive steps:

- 132 1. Differences of subjective and some “latent”, “objective”, or “unified” party
133 placements constitute a necessary but not sufficient condition for the pres-
134 ence and detection of projection effects. Therefore we compute the deviation
135 of voter-specific party placements from some baseline measure of “true”
136 positions.
- 137 2. A sufficient condition for the definition of projection bias exists if voters
138 deflate (inflate) spatial distances for candidates or parties they like (oppose).
139 To measure whether this condition is met, we relate the calculated placement
140 differences to some measure of party attachment and evaluate the marginal
141 effects as a sufficient condition indicating projection effects.

142 2.1 Objective Party Positions

143 A crucial point for the measurement of projection effects is whether we need
144 to assume some yardstick of objective party positions and, if so, how this refer-
145 ence may be theoretically justified and empirically assessed. Of course, the very
146 notions of objective or true party positions are conceptually dubious, and guess-
147 ing “where a party really stands” intimates somewhat esoteric assumptions and
148 goals. Measurement strategies suggested by political scientists are often ambigu-
149 ous. Established measures of party ideology or preferences frequently start with
150 incompatible assumptions, are usually designed for different purposes, and differ
151 in their empirical assessments.

152 In empirical terms, the studies by Granberg and Holmberg (1986), Granberg
153 et al. (1988), Merrill et al. (2001), and Adams et al. (2005) reach very similar
154 conclusions. There is ample, solid, and robust evidence for the existence of pro-
155 jection effects. Its two components, assimilation and contrast, appear to be of
156 approximately similar significance in U.S. presidential and congressional races.
157 Beyond the United States, in multiparty systems, the assimilation of preferred
158 parties seems to have a much more significant effect than the contrast of disliked
159 ones. This has often been called the asymmetry of projection effects, which is
160 governed by the alleged dominance of assimilation over contrast effects (Granberg
161 and Holmberg 1986; Granberg et al. 1988; Krosnick 2002; Merrill et al. 2001).

2.2 Placement Differences and Projection Bias

A necessary condition for the assertion of projection effects is a difference between spatial distances based on subjective and objective party placements. Considering that our unit of analysis is not the location voter i assigns to party j , but rather the distance between the voter's self-placement and the assigned spatial party position, our measurement model may be re-phrased and focused on the difference of the two utility specifications U and U^* :

1. Subjective distances/utilities ΔU are given by the distance of the voter's self-placement (v_i) and subjective, voter-specific party placements p_{ij} ($U = -|v_i - p_{i,j}|$).
2. objective distances/utilities U^* are defined by the negative distance of the voter's self-placement and objective, unified party positions p^* ($U^* = -|v_i - p_{j^*}|$).
3. the difference of these distances, i.e., the difference of subjective and objective placement distances ΔU , is, in turn, given by

$$\Delta U = U - U^* = -|v_i - p_{ij}| + |v_i - p_{j^*}|$$

Note that we label the difference between voter-specific and mean party locations (p_{ij} and p_{j^*}) relative to the voter v_i placement differences with ΔU . Actual values of ΔU yield some first insight into potential projection effects and differentiate between three alternative scenarios:

1. $\Delta U = 0$: if ΔU equals 0, the voter's assessment of a political party equals the true position, and projection bias is absent by definition,
2. $\Delta U > 0$: if ΔU is positive, subjective party placements yield a higher utility than mean party positions, and this *may* indicate projection bias (assimilation),
3. $\Delta U < 0$: if ΔU is negative, utilities based on objective placements exceed subjective ones, and this *may* also point to projection bias (contrast).

At this point, we hasten to add one additional caveat: there are a number of cases where the voter does not only pull a party *towards* her own position but even pulls her idiosyncratically perceived party position *beyond* her personal ideal point, thereby reversing the ordering of party and voter positions ("leapfrogging"). For conceptual reasons, we have decided to remove all cases from our sample where

194 the party position did not remain on the same ‘side’ either on the left–right
195 dimension.

196 There are many potential reasons why voter-specific party placements may
197 differ from objective positions of political parties, and projection bias, be it assim-
198 ilation or contrast, is only one possible explanation. The formal identification of
199 projection bias requires to relate the placement differences ΔU to some measure
200 of party attachment and to compute the marginal effect of party identification on
201 placement differences. Let us therefore assume that we obtain an indicator A for
202 the attachment of the voter to any of the parties contesting in an election. Lower
203 (higher) values of A indicate that she opposes (likes) this specific party.

$$204 \quad \Delta U \propto A$$

$$205 \quad \Delta U = \beta A + \beta_0$$

206 If there is projection bias, placement differences ΔP and party attachment A
207 should be positively related: $\beta_1 > 0$: $\Delta U = \beta_1 A_{ij} + \beta_0$. The corresponding
208 marginal effect is given by $\partial \Delta U / \partial A_{ij} = \beta_1$. Projection effects are only present
209 when $\beta_1 > 0$. The intersection of the upward-sloping predicted effects with the
210 zero-line ($\Delta U = 0$, $U = U^*$) by definition separates contrast and assimila-
211 tion. The presence and magnitude of projection effects are thus indicated by
212 the marginal effects of party attachment A on the difference of objective and
213 subjective placement differentials or utilities ΔU . In the subsequent section, we
214 will discuss specific options to measure party attachment, explore the functional
215 form of the effects of party attachment on placement differentials, and suggest a
216 context-sensitive, comparative measurement model.

217 **3 Measuring Projection Bias**

218 The empirical section begins with a brief introduction of the data at hand and
219 a presentation of the key decisions taken to operationalize and measure the key
220 variables. The subsequent presentation of empirical findings focuses on the pres-
221 ence and magnitude of projection effects, critically reviews some specification
222 issues, and addresses both the presence and variability of projection effects. We
223 conclude that assimilation and contrast are clearly visible, statistically significant,
224 and potentially consequential effects, but we also demonstrate that their impacts
225 vary considerably across individual parliamentary elections and across alternative
226 political parties nested in these contexts.

227 In this paper, we present a secondary analysis of the “Comparative Study
 228 of Electoral Systems” series. CSES is a collaborative research project which is
 229 organized by survey teams from a growing number of countries that field common
 230 modules of questions in their respective post-election surveys. The CSES project
 231 thus aims at providing a unified infrastructure for the comparative and contextual
 232 study of electoral politics. To date, the series consists of four complete modules,
 233 with each covering some forty to fifty country-election segments.

234 To date, the parallel election surveys gathered by the CSES project are *the*
 235 unique and exhaustive data source for comparative electoral studies. Our statisti-
 236 cal models of vote choice build upon the “CSES Integrated Module Dataset”
 237 (henceforth: CSES IMD dataset), i.e., upon a cumulation of the first four waves
 238 of CSES which covers a heterogeneous set of more than 280,000 interviewees,
 239 174 individual elections, and 55 countries. The cumulative dataset used for the
 240 subsequent analyses covers an overall period from 1996 to 2016. For consis-
 241 tency, we exclusively focus on elections to national parliaments, to lower houses
 242 in bicameral systems. The dataset compiled for our analyses includes of a subset
 243 of 148 post-election surveys from 53 different, heterogeneous polities. Most of
 244 the election segments cover interviews with roughly 1000 to 2000 eligible voters,
 245 with a minimum of $N = 860$ voters in Hong Kong (2004) and a maximum of $N = 4495$
 246 voters in Canada (2008). (Downloadable data, further details about the
 247 aims, the components, and the methodological foundations of the CSES project
 248 can be assessed on the internet at: <http://www.cses.org/>.)

249 In the next step, we discuss the subset of variables selected for the comparative
 250 assessment of projection effects. First, we introduce variables needed to establish
 251 the necessary condition for projection effects. The CSES questionnaires explore
 252 voter self-placements on the general left–right scale ($v_i \in [0, 10]$) and also require
 253 the respondents to place each party competing on the same general left–right
 254 dimension ($p_{i,j} \in [0, 10]$). Subjective distances P are based on voter-specific
 255 party placements ($U = |v_i - p_{i,j}|$) and objective distances are derived from mean
 256 voter-specific party placements ($U* = |v_i - p_j^*|$). The placement difference
 257 is given by subtracting the objective utilities from the subjective ones: $\Delta U =$
 258 $U - U* = -|v_i - p_{i,j}| + |v_i - p_j^*|$.

259 Secondly, we turn to an additional variable that enables us to establish suf-
 260 ficient conditions for projection effects. Alongside the key variables indicating
 261 the self-reported spatial positions of voters and their perceived party placements,
 262 the CSES questionnaires also include a fine-grained thermometer indicator for
 263 party evaluation. Respondents are inquired to indicate whether they like a party
 264 or not on a scale ranging from 0 [“strongly dislike”] to 10 [“strongly like”]. In the
 265 English language questionnaires, the specific wording is: “I’d like to know what

266 you think about each of our political parties. After I read the name of a political
267 party, please rate it on a scale from 0 to 10, where 0 means you strongly dislike
268 that party and 10 means that you strongly like that party". The empirical distribu-
269 tion of thermometer scores is considerably skewed to the left. Significantly, more
270 than forty percent of respondents, when asked whether they would like a specific
271 political party, promptly indicate that they do "strongly dislike" it (with a score
272 at or close to zero).

273 Applying an exact, graded measure of party attachment provides additional
274 information. It goes beyond preceding studies that persistently employed some
275 binary distinction between voters and non-voters of a specific candidate or party.
276 This gross measure may well be sufficient for the analysis of two-party com-
277 petition in U.S. presidential and congressional elections. However, in multiparty
278 systems, we believe this oversimplifies the more complex and graduated pref-
279 erence structures. When there are multiple political parties competing, lumping
280 a variety of different voters, who possibly hold very different political prefer-
281 ences, into the broad and diffuse category of "non-supporters" does not yield
282 sufficient analytical accuracy and fails to fully exploit the rich data at hand. This
283 has the potential to introduce bias in our measurement and is likely to decrease
284 the empirical significance of contrast effects.

285 4 The Magnitude of Assimilation and Contrast

286 We evaluate the presence and magnitude of projection effects by applying hier-
287 archical linear models which consider the election, the party, and the individual
288 level, and we organize our presentation of the empirical finding in two consec-
289 utive steps: (i) The fixed part of the hierarchical models sheds some light on
290 the presence and the magnitude of projection effects. (ii) The random part of
291 the measurement model enables us to gauge the variability of assimilation and
292 contrast across different elections and different parties competing.

293 Remember that we establish assimilation whenever utilities based on sub-
294 jective, voter-specific party placements exceed objective utilities, which are
295 calculated with mean party positions ($U > U^*$). In turn, there are contrast effects
296 whenever objective utilities exceed subjective utilities ($U^* > U$). So as to iden-
297 tify projection effects, we relate these placement differences ΔU to our measure
298 of closeness to a particular political party, being the thermometer scale of party
299 evaluations (LIKE $\in [0, 10]$), higher values of LIKE indicate more favorable
300 evaluations and thus a higher attachment to a political party. We identify projec-
301 tion effects whenever placement mismatch ΔU and closeness to political parties

302 LIKE_{*i,j*[*k*]} are related positively. The measurement model accounts for placement
 303 differences ΔU by hierarchical linear models, which probe variation across differ-
 304 ent countries and parties. We may formalize and present a Bayesian hierarchical
 305 model as follows:

$$306 \quad \Delta U_{i,j[k]} \sim N(\text{LIKE}_{i,j[k]} \boldsymbol{\beta}_{[k]}, \sigma_{[k]}^2)$$

$$308 \quad \boldsymbol{\beta}_{[k]} \sim N(\mathbf{z}_{j[k]} \boldsymbol{\gamma}_{\mathbf{k}}, \omega_{\mathbf{k}}^2)$$

309 There are two alternative ways to estimate the measurement model. LIKE_{*i,j*[*k*]},
 310 which is captured by an eleven-point-scale, could either be treated as a categorical
 311 or as an interval variable. Treating LIKE_{*i,j*[*k*]} as a discrete variable and introduc-
 312 ing a series of dummy variables for the levels of LIKE_{*i,j*[*k*]} yields a somewhat
 313 more complex model but avoids some dubious and unverified assumptions concern-
 314 ing the effect of LIKE_{*i,j*[*k*]} on ΔU . Treating LIKE_{*i,j*[*k*]} as a continuous
 315 variable is more convenient due to the ease of statistical estimation and the inclu-
 316 sion of interaction terms. However, this strategy involves unverified assumptions
 317 about the linearity of parameters that are neither supported by theoretical models
 318 nor backed by available empirical evidence.

319 We first discuss a hierarchical linear model that treats LIKE_{*i,j*[*k*]} as a cate-
 320 gorical predictor and employs random effects on the country and party levels.
 321 Breaking down the party attachment variable into a series of eleven binary
 322 dummies enables us to estimate its effect on assimilation and contrast without
 323 assuming any specific (linear) association. This specification has the obvious
 324 advantage that it does not assume any specific functional association between
 325 the level of closeness to a certain party and the magnitude of assimilation and
 326 contrast. The fixed part of the multilevel model is utilized to explore the aver-
 327 age magnitude of assimilation and contrast in the stacked CSES dataset. The
 328 empirical estimates confirm highly significant effects of LIKE_{*i,j*[*k*]} on placement
 329 mismatch and refer to meaningful and substantive projection effects in voter-
 330 specific party placements. This (and the following) models are estimated using
 331 uninformative priors for $\boldsymbol{\beta}$ and weakly informative priors for σ^2 with mean set
 332 equal to the marginal variance of the dependent variable. We use 10,000 itera-
 333 tions of the MCMC algorithm, a burnin of $N = 1000$, one Markov chain and no
 334 thinning to derive the posterior distributions of our key parameters ($\boldsymbol{\beta}$ and $\widehat{\Delta U}$).
 335 The hierarchical models pass all usual convergence and specification tests, and
 336 the Markov chains display quick mixing.

337 Table 1 summarizes the posterior distributions of the eleven categorical predic-
 338 tors and thus provides an overview of baseline assimilation and contrast effects

Table 1 Posterior summaries with $\text{LIKE}_{i,j[k]}$ as a categorical predictor

Party Attach	Distribution		Quantiles	
$\text{LIKE}_{i,j[k]}$	β	σ	2.5%	97.5%
0	-1.65	0.02	-1.69	-1.62
1	-1.14	0.02	-1.18	-1.10
2	-0.83	0.02	-0.87	-0.79
3	-0.57	0.02	-0.61	-0.53
4	-0.35	0.02	-0.39	-0.31
5	-0.18	0.02	-0.22	-0.15
6	-0.06	0.02	-0.10	-0.02
7	0.02	0.02	-0.02	0.06
8	0.13	0.02	0.09	0.16
9	0.23	0.02	0.19	0.27
10	0.59	0.02	0.55	0.63
N	Level 1: 713,439 individuals/ Level 2: 148 election studies			

Notes: The Table displays marginal effects over the different categories of $\text{LIKE}_{i,j[k]}$ derived from a Bayesian hierarchical model. The dependent variable captures the difference of utilities based on voter-specific and mean party placements on the left–right dimension (ΔU). i indicates the individual, j indicates the party, and k indexes the election level. The dataset is organized in a two-level structure with 713,439 complete and valid respondents nested within the contexts of exactly 148 elections compiled from the CSES IMD dataset. Random intercepts and random slopes for all categories of $\text{LIKE}_{i,j[k]}$ are included at the election level, but have been omitted in the tabular summary

339 across the selected country segments of cumulated CSES dataset. The results
 340 yield strong evidence on the significance of projection effects and clearly demon-
 341 strate that the magnitude of assimilation and contrast clearly increases/decreases
 342 with the levels of party attachment captured by the $\text{LIKE}_{i,j[k]}$ indicator. Every-
 343 thing else being equal, a voter tends to pull parties she strongly prefers almost
 344 0.6 scale points toward her personal position ($\text{LIKE}_{i,j[k]} = 10$). On the other
 345 hand, she tends to push parties she strongly opposes more than 1.6 scale points
 346 away ($\text{LIKE}_{i,j[k]} = 0$).

347 We next turn to a more parsimonious multilevel regression model that treats
 348 $\text{LIKE}_{i,j[k]}$ as a continuous predictor by introducing some additional assumptions
 349 regarding the shape of the association between ΔU and LIKE. The previous
 350 model indicated non-linearities of projection effects, and considering that contrast
 351 effects appear to be much stronger than assimilation, we have decided to use

Table 2 Posterior summaries with $\text{LIKE}_{i,j[k]}$ as a continuous predictor

Party Attach	Distribution		Quantiles	
	β	σ	2.5%	97.5%
β_1 : [$\lg \text{LIKE}_{i,j[k]}$]	0.91	0.03	0.85	0.96
β_0 : [intercept]	-1.80	0.06	-1.91	-1.68
Random Effects				
$\zeta_{0[k]}/\delta[\beta_0]$	0.70	0.04	0.63	0.79
$\zeta_{1[k]}/\delta[\beta_1]$	0.35	0.02	0.31	0.39
N	Level 1: 713,439 individuals/ Level 2: 148 election studies			

Notes: The Table displays marginal effects over logged values of the party attachment indicator $\text{LIKE}_{i,j[k]}$ taken from a Bayesian hierarchical model. Random intercepts and a random slope $\lg \text{LIKE}_{i,j[k]}$ are included at the election level but have been omitted in the tabular summary

the natural logs of $\text{LIKE}_{i,j[k]}$ as our key independent variable. (Given that the original $\text{LIKE}_{i,j[k]}$ scale ranges from zero to ten and that the logarithm of zero is not defined, we have shifted the scale by adding one point. The logged scale thus may take on values from $\ln 1$ to $\ln 11$.) The hierarchical measurement model includes random intercepts and random slopes at the election level.

Table 2 presents a hierarchical model that confirms the presence of projection effects across the countries covered in the three CSES waves. The multilevel model rests on rich empirical data provided by 713,439 “complete” observations compiled from 148 post-election surveys covered by the CSES project, which, in turn, are taken from 53 different country contexts. The empirical findings clearly reveal the expected positive effect of logged thermometer scores on the difference of subjective and objective utilities. The obtained random effects, noted in the lower part of the table, also demonstrate that the magnitude of assimilation and contrast varies, i.e., the estimated random intercepts and random slopes, considerably vary across different contexts.

Figure 1 illustrates our inferences from the two alternative measurement models graphically and provides a comparison of the previous specifications, which capture the magnitude of assimilation and contrast and contrast depending on party attachment. We concentrate on the “fixed part” of the models. The left-hand panel displays violin plots of posterior densities based on models which treat party attachment ($\text{LIKE}_{i,j[k]}$) as a categorical variable (cf. Table 1), and the right-hand panel displays similar graphs when $\text{LIKE}_{i,j[k]}$ is included as a logged

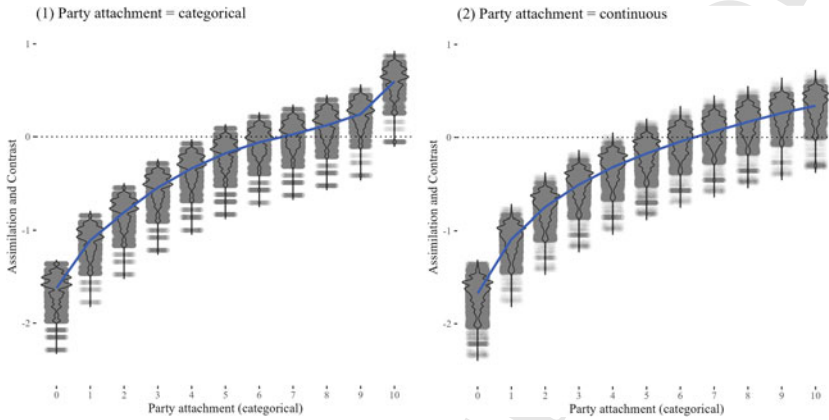


Fig. 1 Comparing Inferences and Models of Projection Effects (Fixed Part) **Notes:** The Figure reproduces the mean trajectories for projection bias on the left–right dimension. 1) The left-hand panel shows predictions of the magnitude of assimilation and/or contrast based on the “categorical” model, 2) the right-hand panel presents similar predictions for the “continuous” model. In either panel, the dashed horizontal line ($\Delta P = 0$) separates contrast (below) from assimilation effects (above)
Source: Own graphic

374 continuous predictor (cf. Table 2). Since both predictions are almost congru-
 375 ent, our confidence in the descriptive inferences and the specification of *logged*
 376 $\text{LIKE}_{i,j[k]}$ as the key independent variable is soundly corroborated.

377 Utilizing a more fine-grained measure of party attachment, empirical find-
 378 ings derived from both models refute the established asymmetry hypothesis.
 379 In contrast to key works of dissonance (Festinger 1957) or congruity theory
 380 (Osgood and Tannenbaum 1995) and also in contrast with previous work con-
 381 cerned with bipartism in the U.S. (Granberg and Holmberg 1986; Granberg et al.
 382 1988; Krosnick 2002), the average consequences of contrast clearly and sys-
 383 tematically outweigh the average effects of assimilation. We will return to this
 384 modified perspective on the “asymmetry” of projection effects when discussing
 385 the implications of assimilation and contrast for (models of) vote choice in more
 386 detail.

387 Note that the overall projection effect $\beta_{1[k]}$ is composed of the fixed (β_1) and
 388 the random parts ($\zeta_{1[k]}$) of the hierarchical model. While the previous Tables and
 389 Figures have summarized the fixed part of the Bayesian measurement model, we
 390 now explore the context-dependency and variation of projection effects across 148

391 parliamentary elections selected from the first three waves of the CSES project.
392 As indicated, the measurement model captures the contextual variation of assimilation
393 and contrast by the random components on $\text{LIKE}_{i,j[k]}$ across the various
394 electoral contexts. Figure 2 displays caterpillar plots that capture (the variation
395 of) posterior densities of $\zeta_{1[k]}$. Projection effects appear to be particularly strong,
396 for instance, in the U.S. elections of 2004, the South African elections of 2009,
397 and the Taiwanese elections of 2001, while assimilation and contrast are almost
398 absent or very weak in Bulgaria 2001, Chile 2009, Brazil 2006, or the Walloon
399 part of Belgium in 1999.

400 5 Correcting for Projection Bias

401 Our theoretical assumptions have been matched by rich empirical evidence, refer
402 to the significance, and illustrate the magnitude of projection effects. Moreover,
403 the universality of the assimilation and contrast logics across the wide range of
404 countries and parliamentary elections illustrates the importance of considering
405 these distortions in models of vote choice. The level and magnitude of assimilation
406 and contrast effects appear to be alarming as well: if a voter on average pulls
407 a party she likes by more than 0.5 scale points towards and pushes a party she
408 dislikes by more than 1.6 scale points away from her personal ideal point, pro-
409 jection effects may not only alter the raw values of distance terms but potentially
410 modify the rank ordering of spatial utility terms.

411 This section builds upon the previous theoretical and empirical knowledge
412 and systematically explores the consequences of projection bias for party eval-
413 uation and vote choice. We first demonstrate how the predictions for the bias
414 of projection effects may be utilized to “correct” and rescale idiosyncratic
415 party placements. Subsequently, we provide comparisons of spatial voting mod-
416 els which build upon i) subjective, unscaled idiosyncratic party placements, ii)
417 objective, mean party placements, and iii) corrected, rescaled idiosyncratic party
418 placements.

419 5.1 Unbiased Party Placements

420 The spatial model of voting is often considered the workhorse of modern elec-
421 toral studies. Downsian proximity models posit that vote choice is determined by
422 ideological and programmatic proximity and distance (cf., for instance, Downs
423 1957; Westholm 1997). Both voters and parties are thus represented by points

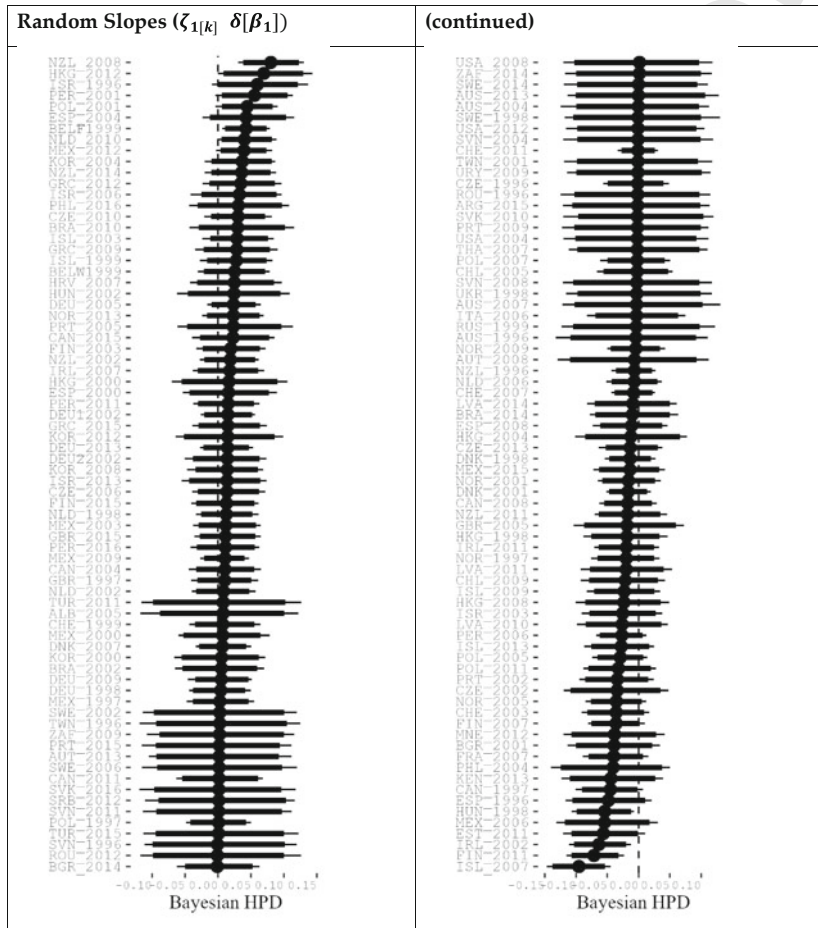


Fig. 2 Projection Effects Across Heterogeneous Electoral Contexts **Notes:** The caterpillar plots present the posterior densities of the election-specific random components $\zeta_{1[k]}$ on $\lg \text{LIKE}_{i,j[k]}$. Empirical random effects and their distributions are obtained from the Bayesian measurement model presented in Table 2.

Source: Own graphic

424 within the political space. Voters evaluate party alternatives by spatial proximity,
 425 and thus the candidate or party which is closest to the voter's preferences in an
 426 n -dimensional political space yields the highest utility. The principle of utility
 427 maximization implies that any voter v_i prefers the electoral platform that offers
 428 her the highest expected utility.

429 We employ a risk-neutral metric for the measurement of spatial utility, i.e., we
 430 assess ideological proximity or distance on the left–right scale by the absolute
 431 distance of voter ideal points v_i and party placements $p_{i,j}$. Essentially, idiosyn-
 432 cratic utilities are defined by $U_{i,j[k]} = -|v_{i[k]} - p_{i,j[k]}|$, objective utilities are
 433 given by $U_{i,j[k]} = -|v_{i[k]} - p^*_{j[k]}|$, and corrected, rescaled utilities are provided
 434 by $U_{i,j[k]} = -|v_{i[k]} - p'_{i,j[k]}|$. From another formal angle, projection effects
 435 may also be assessed by the difference of two absolute distances, which may be
 436 conveniently rewritten as the difference of subjective and objective utility terms:

$$437 \quad \Delta U_{i,j[k]} = -|v_{i[k]} - p_{i,j[k]}| - |v_{i[k]} - p^*_{j[k]}|$$

$$438 \quad = U_{i,j[k]} - U^*_{j[k]}$$

440 Scaled, unbiased utilities are now derived by subtracting predicted projection bias
 441 $\widehat{\Delta U}$ from unscaled utility terms. If $\widehat{\Delta U} > 0$ the model indicates assimilation, and
 442 rescaled utilities are lower than unscaled utilities, if $\widehat{\Delta U} < 0$, the model refers to
 443 contrast, and rescaled exceed unscaled utility terms.

444 Our unified model also includes a vector of non-spatial variables $t_{i,j[k]}$ so that
 445 the overall utility $U_{i,j[k]}$ of party \mathbf{P} for voter \mathbf{V} amounts to:

$$448 \quad U_{i,j[k]} = -\|\mathbf{V} - \mathbf{P}\| = \sum_{a[k]=1}^n \alpha_{a[k]} U_{i,j[k]} + \beta \mathbf{t}_{i,j[k]}$$

448 Given that we have presented ample evidence on the genuineness of projection
 449 effects in idiosyncratic party placements $p_{i,j[a]}$, the provision of assimilation and
 450 contrast may potentially inflate estimates of the “spatial” coefficients α and may
 451 potentially depress those of non-spatial predictors β . We address these issues by
 452 replacing utility terms based on idiosyncratic party placements $U_{i,j[k]}$ by utility
 453 terms based on corrected, unbiased party placements $U'_{i,j[k]}$ derived from the
 454 posterior distribution of the measurement model:

$$455 \quad U^*_{i,j[k]} = \alpha_k U'_{i,j[k]} + \beta \mathbf{t}_{i,j[k]}$$

$$456 \quad = \alpha_k (U_{i,j[k]} - \widehat{\Delta P}) + \beta \mathbf{t}_{i,j[k]}$$

$$= \alpha_k \left(|v_{i[k]} - p_{i,j[k]}| - \widehat{\Delta P} \right) + \beta \mathbf{t}_{i,j[k]}$$

5.2 An Application to the Cumulated CSES Data

We now discuss the actual consequences of projection bias for the estimation of policy (α) and non-spatial (β) predictors of vote choice. Our unified models of vote choice represent the spatial component by classical Downsian proximity models, which are either based on i) idiosyncratic, ii) on mean, or iii) on corrected, rescaled party placements. Non-spatial determinants of the vote are represented by a binary variable which indicates party identification, i.e., whether the interviewee declares to be “close to” a specific party alternative or not, and we also include alternative-specific intercepts in order to capture unmodeled utility components (sometimes labeled “valence”). The discrete choice models are estimated separately for each election by conditional logit:

$$\Pr[v_{i[k]} = j | u_{i,j[k]}, \mathbf{t}_{i,j[k]}] = \frac{e^{u_{i,j[k]}}}{\sum_{j=1}^J e^{u_{i,j[k]}}} = \frac{e^{\alpha |v_{i[k]} - p'_{i,j[k]}| + \beta \mathbf{t}_{i,j[k]}}}{\sum_{j=1}^J e^{\alpha |v_{i[k]} - p'_{i,j[k]}| + \beta \mathbf{t}_{i,j[k]}}}$$

The models are estimated using post-election survey data from the cumulated CSES IMD dataset. The cumulated CSES IMD dataset provides all necessary items. The dependent variable is reported vote choice. Turning to the independent variables, the respondents are inquired to indicate their ideological self-placements v_i on an eleven-point scale: “where would you place yourself on a scale from 0 to 10 where 0 means the left and 10 means the right”. Similarly, all voters are required to place all party alternatives on a similar scale. Eventually, non-spatial considerations $t_{i,j[k]}$ are captured by party identification: “Do you usually think of yourself as close to any particular party?” and, if yes, “Which party do you feel closest to?”.

A fundamental interest in the comparison of parameters is to detect whether biased, unscaled party placements effectively introduce an upward bias to spatial (α_k) and a downward bias to non-spatial (β_k) determinants of vote choice. The cumulation of the first four waves of the CSES project provides us with a huge dataset, which has been exploited to assess the presence and magnitude of projection effects. However, our empirical comparison of alternative measures in models of vote choice models can only focus on a subset of the rich CSES data.

489 The utilization of election-wise conditional logit models implies discarding sur-
 490 vey respondents that did not place all party alternatives on the left–right scale.
 491 Further, party identification, one of the critical variables that capture non-spatial
 492 motives of vote choice, has not been included in many country election studies,
 493 and these segments of the CSES data will also be excluded.

494 We begin our presentation of the results by comparing election-specific condi-
 495 tional logit models based on i) subjective, unscaled and on ii) objective, mean
 496 placements of the party alternatives ($p_{i,j[k]}$ and $p^*_{j[k]} = \overline{p_{i,j[k]}}$, respectively).
 497 Both relying on idiosyncratic and on mean placements of the candidates and par-
 498 ties competing, the coefficients on Downsian proximity utilities are, as expected,
 499 positive, precisely estimated and substantially meaningful in the remaining 148
 500 election segments.

501 Our key interest is to explore whether the presence of substantively mean-
 502 ingful assimilation and contrast effects also introduces systematic bias to the
 503 coefficients obtained for both the spatial and non-spatial components of voter
 504 utility. To explore this in more detail, Fig. 3 provides a comparison of coeffi-
 505 cient estimates, which are obtained from a series of conditional logit models and

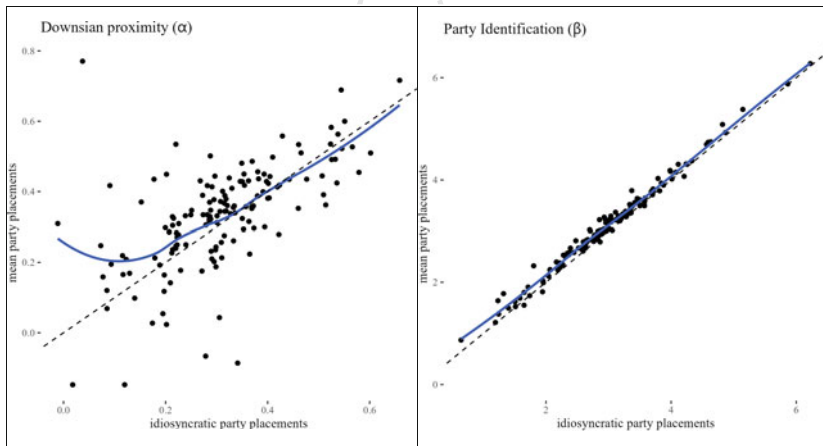


Fig. 3 Unscaled and Mean Placements, Spatial, and Non-Spatial Utilities **Notes:** The left-hand panel compares coefficient estimates on the spatial, the right-hand panel those on the non-spatial impacts on vote choice. In either panel, the association of parameter estimates is indicated by a linear regression line and a LOESS smoother.

Source: Own graphic

506 which are based upon either idiosyncratic and on mean placements. The left-hand
507 panel displays coefficient estimates on spatial proximity terms (α), the right-hand
508 panel coefficients on party identification (β). In either panel, estimates based on
509 idiosyncratic party placements are displayed on the x -axis, and parallel findings
510 based upon mean party placements are shown on the y -axis.

511 The left-hand panel reveals that model parameters are affected by choice of
512 either idiosyncratic or mean party placements. Comparing both specifications over
513 the 148 election segments in the dataset, estimates on the coefficients $\alpha_{[k]}$ are
514 associated, but there is also some considerable random noise. For some of the
515 components, for instance the parliamentary elections in Italy (2004), South Africa
516 (2009), and Spain (2006), these differences are large and substantial. The level
517 and magnitude of the coefficients $\alpha_{[k]}$ also indicates the impact and salience of
518 spatial voting for electoral behavior. In contrast to previous assumptions, how-
519 ever, the utilization of idiosyncratic instead of mean party placements does not
520 systematically inflate estimates on the relevance of spatial proximity in the data at
521 hand. Rather, the differences produced by the alternative measurement strategies
522 appear to be unsystematic. Measuring party positions by idiosyncratic instead of
523 objective yardsticks does not systematically inflate the relevance of spatial voting.
524 These findings underline the inferred salience of spatial voting only rarely may
525 be systematically distorted by assimilation and contrast.

526 The right-hand panel illustrates that these discrepancies do not carry over to
527 the non-spatial predictors $\beta_{[k]}$ in the unified models. Regardless whether party
528 positions are assessed via idiosyncratic or via mean placements, the estimated
529 impact of party identification on vote choice is almost identical and overlaps
530 with only very limited random noise.

531 Summing this up, the comparison of models based on idiosyncratic and mean
532 placements does not corroborate the expectations of artificially inflated coeffi-
533 cients on spatial proximity and artificially depressed coefficients on non-spatial
534 or valence components of the utility. Instead, empirical findings derived with
535 raw and with mean party placements correlate significantly, while predictors for
536 non-spatial terms are essentially identical.

537 Figure 4 provides parallel comparisons of coefficient estimates based upon
538 idiosyncratic and corrected, re-scaled party placements. As above, the left-hand
539 panel displays coefficient estimates on spatial proximity terms (α), the right-hand
540 panel coefficients on party identification (β). The left-hand panel reveals that
541 coefficients derived from raw, idiosyncratic and from corrected, rescaled utility
542 terms correlate very strongly. With “raw” estimates shown on the x - and rescaled
543 coefficients shown on the y -axis, we find that all points in are located below the
544 dashed identify line. Thus, coefficients derived from raw party placements are

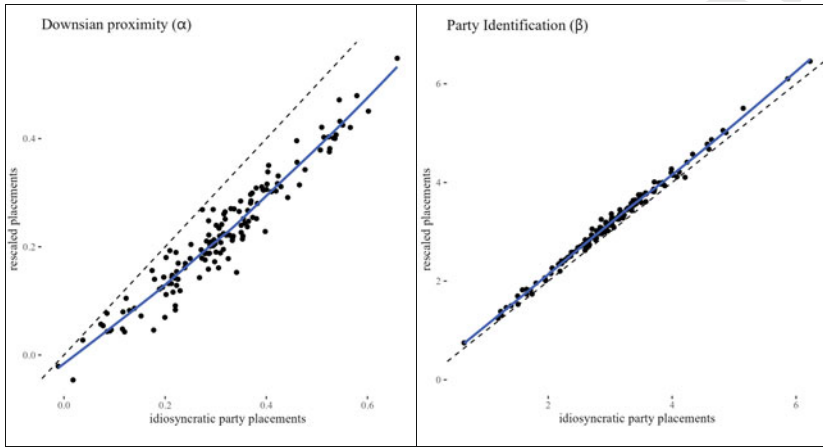


Fig. 4 Unscaled and Scaled Placements, Spatial, and Non-Spatial Utilities **Notes:** The left-hand panel compares coefficient estimates on the spatial, the right-hand panel those on the non-spatial impacts on vote choice. In both panels, i) the association of the parameter estimates is indicated by a linear regression line and a LOESS smoother.

Source: Own graphic

545 systematically inflated, while coefficients obtained from rescaled placements are
 546 consistently lower. In other words: while findings from raw and from rescaled
 547 party placements continue to correspond closely and systematically, the graphical
 548 representation of our comparative findings clearly reveals systematic bias due to
 549 idiosyncratic party placements and projection effects. Ultimately, the right-hand
 550 panel of Fig. 4 illustrates that measurement bias in the spatial utility component
 551 does not carry over to biased inferences concerning the non-spatial utility of vote
 552 choice.

553 On the one hand, we have gathered empirical evidence for substantial levels of
 554 assimilation and contrast in idiosyncratic party placements. On the other hand, we
 555 also conclude that measurement bias only partially carries over to estimation bias
 556 in empirical models of vote choice. Across the 148 election studies covered by the
 557 CSES integrated module dataset, assimilation and contrast tend to systematically
 558 inflate the salience parameters attributed to spatial utility components. However,
 559 these distortions are not too large and confined to spatial utility terms. Vice versa,
 560 inflated spatial utility terms do not yield biased estimates of non-spatial utility.

561 Part of an explanation may be the detected asymmetry of projection effects,
 562 i.e., that the magnitude of contrast (more than 1.6 scale points) across the board

563 appears to be much larger than the magnitude of assimilation effects (more than
564 0.5 scale points). Given that voters rarely choose candidates or parties which
565 are both distant in spatial terms and unpopular in valence terms, assimilation
566 rather than contrast effects determine and/ or change vote choice. Because these
567 empirical assimilation effects are small, the implications of projection effects on
568 vote choice may, across the board, not be as significant as suggested in some
569 conceptual and theoretical contributions.

570 **6 Findings, Remedies, and Perspectives**

571 The take-home messages of this paper are two-fold. First, in descriptive terms,
572 idiosyncratic spatial party placements are systematically contaminated and dis-
573 torted by the non-spatial motives of the vote: voters tend to pull candidates or
574 parties they like for non-spatial reasons systematically towards their ideal points,
575 and the same voters tend to push candidate or parties they dislike systematically
576 away from their own positions. Empirical projection effects are of a considerable
577 and meaningful magnitude but also vary substantially over the heterogeneous
578 national election contexts. Projection effects are also asymmetric, in contrast to
579 some previous studies, our analysis clearly underscores that the magnitude of
580 contrast effects outweighs the strength of assimilation effects.

581 Secondly, while projection effects are an interesting phenomenon per se, there
582 could also be significant implications for models of electoral choice and place-
583 ment bias could also be reflected in estimation bias regarding key parameters
584 in spatial and in unified models of vote choice. Specifically, many researchers
585 concerned about projection effects and their consequences have suggested that
586 assimilation and contrast may artificially inflate the salience of spatial proximi-
587 ty and artificially depress the importance of alternative utility terms such as
588 directional voting or of non-spatial predictors such as valence (however oper-
589 ationalized) or party identification (MacDonald et al., 1991, 1995, 1998,2001,
590 2007; Rabinowitz and MacDonald 1989).

591 We have addressed these issues by developing a rescaling procedure based
592 upon the results of our hierarchical projection effects model. However, a com-
593 parison of biased, idiosyncratic utility terms and unbiased, rescaled utility terms
594 did not affirm the above concerns. These different specifications do enable us to
595 detect some systematic bias in models of vote choice which arise due to projec-
596 tion effects. However, these differences are usually modest, projection effects tend
597 to inflate the salience of proximity voting, but do not carry over to assessments

598 of non-spatial utility. Additional robustness checks based on mean party place-
599 ments did not indicate any systematic upward or downward bias in the obtained
600 coefficients but rather highlights the sensibility of the discrete choice models to
601 the introduction of measurement error by mean party placements.

602 Further empirical exploration is required to verify these results and to establish
603 why distorted party placements do not mechanically carry over to yield system-
604 atically biased assessments of projection effects. We believe that at least part of
605 the explanation lies in the asymmetry of its two components. Contrast effects are
606 relatively small but usually do not matter much for vote choice. However, assim-
607 ilation effects may impact on or even change electoral choice, but their magnitude
608 is limited and often negligible.

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Chapter 16

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