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Estimating and exploring the proportions of inter- and intrastate cattle shipments in the United States

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Abstract

Mathematical models are key tools for the development of surveillance, preparedness and response plans for the potential events of emerging and introduced foreign animal diseases. Creating these types of plans requires data; when data are incomplete, mathematical models can help fill in missing information, provided they are informed by the data that are available. In the United States, the most complete national-scale data available on cattle shipments are based on Interstate Certificates of Veterinary Inspection, which track the shipment of cattle between states; data on intrastate cattle shipments are lacking. Here we develop four new datasets on intrastate cat-

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tle shipments in the U.S., including an expert elicitation survey covering 19 states and territories and three state-level brand inspection data sets. The expert elicitation survey provides estimates on the proportion of shipments that travel interstate over multiple regions of the U.S. These survey data also identify differences in shipment patterns between regions, cattle commodity types, and sectors of the cattle industry. These survey data cover more states than any other source of interstate data; however, one limitation of these data is the small number of participating experts in many of the states, only seven of the 19 responding states and territories had a group size of three or larger. The brand data sets include origin and destination information for both intra- and interstate shipments. These data, therefore, also provide detailed information on the proportion of interstate shipments in three Western states, including the temporal and geographic variation in shipments. Because the survey and brand data overlap in the Western U.S., they can be compared. We find that in the Western U.S. the expert estimates of the overall proportion of cattle shipments matched the brand data well. However, the experts estimated that there would be larger differences in beef and dairy shipments than the brand data show. This suggests the cattle industries in the West may be sending similar proportions of commodity specific cattle shipments over state lines. We additionally used the expert survey data to explore how differences in the proportion of interstate shipments can change predictions about cattle shipment patterns using the example of model-guided suggestions for targeted surveillance in Texas. Together these four data sets are the most extensive and geographically comprehensive information to date on intrastate cattle shipments. Additionally, our analyses

on predicted shipment patterns suggest that assumptions about intrastate shipments could have consequences for targeted surveillance.

Keywords: cattle shipment, intrastate shipment, interstate shipment, expert elicitation, brand inspection

1 **Introduction**

2 Surveillance, tracing and response plans are critical aspects of prepared-
3 ness and control for livestock diseases. Previous work has demonstrated
4 that knowledge of livestock shipments is important for understanding dis-
5 ease spread and therefore, for improving the effectiveness of surveillance and
6 outbreak planning and response activities (van Schaik et al., 2002; Green
7 et al., 2006; Ortiz-Pelaez et al., 2006; Kao et al., 2007; Grear et al., 2014;
8 Gorsich et al., 2018). Emerging and re-emerging livestock infections and the
9 potential for an introduced foreign animal disease, require well informed pre-
10 paredness and response plans both in the United States (U.S.) and around
11 the world. Despite this need, there is a limited amount of information on
12 livestock shipments in the U.S. (Buhnerkempe et al., 2013; Lindström et al.,
13 2013), and this is a considerable hindrance to disease preparedness activi-
14 ties. In particular, for the cattle industry in the U.S., within state shipment
15 patterns are not well described.

16 In the U.S., the most extensive data on cattle shipments are the Interstate
17 Certificates of Veterinary Inspection (ICVIs) that record interstate (between-
18 state) shipments of livestock (Buhnerkempe et al., 2013; Portacci et al., 2013;
19 Gorsich et al., 2016). These data have been used to build a national model
20 for cattle shipments, called the United States Animal Movement Model (US-

21 AMM), that can be used to understand general cattle shipment patterns in
22 the U.S. (Buhnerkempe et al., 2013; Lindström et al., 2013) and have also
23 been used to predict movement of at-risk cattle (Grear et al., 2014; Gorsich
24 et al., 2018). USAMM was also coupled with a disease simulation, called the
25 United States Disease Outbreak Simulation (USDOS), to understand the po-
26 tential for pathogen transmission and disease spread via animal shipments
27 at a national-scale (Buhnerkempe et al., 2014). The USAMM model uses in-
28 formation on interstate shipments to estimate the within state patterns, but
29 complete data to inform this process are lacking, and there is uncertainty in
30 the relative contribution of within versus between state movement to disease
31 spread (Lindström et al., 2013). The characterization of intrastate (within-
32 state) shipment patterns and the relative number of shipments that occur
33 within versus between states are key pieces of information for characterizing
34 shipments at the state, regional or national scale.

35 In the majority of U.S. states, intrastate shipments of cattle are not
36 recorded; however, it is generally assumed that the majority of cattle ship-
37 ments occur within states (USDA, 2009). Because there is not a national
38 source of information on intrastate cattle shipments, data describing this
39 process need to be compiled from different sources. Previous studies on
40 cattle shipments have used data compiled from questionnaires and expert
41 opinion to describe intrastate cattle shipments at a local level (Bates et al.,
42 2001; Liu et al., 2012); however, the scale of these studies makes it diffi-
43 cult to extrapolate regional or even state-level patterns. The main source
44 of directly observed data on intrastate shipments are brand inspection data,
45 which some states use when ownership of animals is transferred or when an-

46 imals are shipped. Largely collected in the Western U.S., brand inspection
47 data capture both intrastate shipments and interstate shipments; however,
48 because these are state-level data, the type of shipments tracked, the infor-
49 mation tracked, geographic coverage and the accessibility of the data (i.e.
50 paper versus electronic) vary from state to state. Despite the differences
51 in data accessibility, and the type of data recorded, brand inspection data
52 provide consistently tracked state-level data on intrastate shipments.

53 The brand inspection data provide detailed information on cattle ship-
54 ments traveling within and between states in the Western U.S. Despite the
55 brand inspection data being limited to a subset of states, it most likely pro-
56 vides the best data available on intrastate shipments. The differences in
57 cattle infrastructure and regional management practices in the cattle indus-
58 try make it probable that differences will also be present in shipment patterns
59 across the U.S. Therefore, information gathered from brand inspection data,
60 though invaluable in states where brand inspection is available, may not
61 provide accurate estimates for states in other regions of the U.S. where pro-
62 duction systems can be very different (e.g. many small farms or areas with
63 a predominance of dairy production). To fill these gaps in knowledge, we
64 implement an expert elicitation survey to explore differences in intra- and
65 interstate cattle shipments across the U.S. The comparison between brand
66 inspection data and expert elicitation estimates in the Western U.S. can pro-
67 vide information on the accuracy of expert estimates. We combine the novel
68 survey data with brand inspection data from three Western states (Cali-
69 fornia, Wyoming and Montana), and one market data set from Montana
70 to provide the first regional estimates of intrastate cattle shipments for the

71 U.S. We also use the expert survey data to explore how changing estimates
72 of the proportion of interstate shipments can alter predictions about cattle
73 shipments, and therefore, targeted surveillance of cattle imported to Texas.

74 **Methods**

75 *Expert Elicitation Survey Development and Implementation*

76 The survey was developed and implemented as a modified Delphi group
77 process. This method was chosen because it is the most commonly used
78 survey method in ecology and veterinary epidemiology and could be adapted
79 to the large number of expert groups required for this study (Kuhnert et al.,
80 2005; Gustafson et al., 2010; Kuhnert et al., 2010; Gustafson et al., 2013).
81 The goal of this survey was to develop data on intrastate cattle shipments
82 with good geographic coverage of the continental U.S.

83 Our expert elicitation survey was designed to gather information about
84 the proportion of interstate cattle shipments at the state-level across both
85 the entire cattle industry and different industry subsets. The survey was di-
86 vided into two sections, one for beef and one for dairy, because management
87 practices differ between these commodities and because it was common for
88 experts to have stronger expertise in one commodity. The survey questions
89 asked about shipments of different types of cattle, and shipments traveling
90 to or from different origin and destination types (market, feedlot, etc). The
91 survey was designed with input from subject matter experts on expert elic-
92 itation, and on beef and dairy cattle, respectively. A complete list of the
93 survey questions can be found in Appendix A.

94 Because the survey questions were written at the state-level, multiple

95 groups of state-level experts participated in the survey. We selected ten
96 states to focus on, California, Colorado, Iowa, Minnesota, North Carolina,
97 Nebraska, New York, Pennsylvania, Tennessee, and Texas. These states were
98 selected because they were found in previous work to be important in the
99 cattle shipment network (Gorsich et al., 2016), and they represent major
100 geographic regions in the U.S. The survey targeted cattle experts with deep
101 knowledge of the cattle industry, including cooperative extension professors,
102 state veterinarians, veterinary medical experts, epidemiologists, cattleman’s
103 association leadership, and USDA personnel.

104 Experts were invited to participate in the survey through two routes. The
105 first route of invitation was targeted to the ten focal states. Experts were
106 identified and invited to participate with a letter explaining the survey pro-
107 cess. If the expert was unable or unwilling to participate in the survey, we
108 requested that they suggest another qualified expert. The second route of
109 invitation was more broad and did not specifically target the focal states; a
110 brochure explaining and inviting participation in the survey was sent out to
111 the state veterinarians, veterinary medical experts, and to the United States
112 Animal Health Association and the Agricultural Marketing Service. These
113 organizations and officials receiving the brochure invitation were in a position
114 to identify key state-level experts or are experts in cattle shipments in their
115 own right. All experts who participated in the survey worked in the cattle
116 industry and were in positions that allowed for observation of cattle shipment
117 practices. The survey was administered online through eSurveysPro (esur-
118 veyspro.com). Additional details regarding the design and implementation
119 of the survey can be found in Appendix B.

120 *Expert Elicitation Survey Analysis*

121 For analysis, survey questions were grouped into shipment categories in-
122 cluding, overall, commodity specific (beef or dairy), feeding channel, breeding
123 channel and market shipments (for details on the specific question groupings
124 see Appendix B). One survey question (question 11) was omitted from analy-
125 ses because the responses and comments from the experts indicated multiple
126 interpretations of the question; responses to the other questions did not in-
127 dicate any other questions were subject to misinterpretation.

128 Individual expert estimates were obtained by taking the mean over their
129 responses to the questions in each cattle shipment category analyzed (overall,
130 commodity specific, feeding, or breeding channel and market). State-level es-
131 timates of the proportion of interstate shipments were found by taking the
132 mean of the individual expert estimates from the state. State-level estimates
133 were then aggregated into regional and national-level proportions of inter-
134 state shipments using both the mean and the median number of interstate
135 shipments out of 100 (or number of farms that ship to interstate destina-
136 tions).

137 The national estimate included all contiguous states that responded to
138 the survey. The participating contiguous states were divided into five regions:
139 West, which included California, Idaho, Montana and Nevada; Plains, which
140 included, Colorado, Nebraska, Oklahoma and Texas; Upper Midwest, which
141 included, Iowa, Minnesota and Wisconsin; Northeast, which included, New
142 York and Pennsylvania, and Southeast, which included, Mississippi, North
143 Carolina, Tennessee and Virginia. These regions are loosely based on the
144 USDA Economic Research Service (ERS) farm production regions (Heimlich,

145 2000); however, because not every ERS region had enough representation,
146 multiple regions had to be grouped. Additionally, the mountain region was
147 not contiguous so Idaho, Montana and Nevada were joined with the pacific
148 region state, California, to create the Western region and Colorado was added
149 to the plains region.

150 *Brand Inspection Forms*

151 Brand inspection data was obtained from three states, California (CA),
152 Montana (MT), and Wyoming (WY). Because each state had its own specific
153 requirements for when a brand inspection is required, the data available
154 from these states were not exactly the same (California Department of Food
155 and Agriculture, 2017; Montana.gov Official State Website, 2017; Wyoming
156 Livestock Board, 2017). The CA and WY data sets each contained one year
157 of data (2009 & 2010, respectively). The data set from MT contained three
158 years of data (2009–2011). From MT we also had a data set of shipments
159 originating at markets for one year (2013). The market data set was similar
160 to the brand inspection data sets in that both intra and interstate shipments
161 are tracked; however, in MT, shipments to and from markets were tracked
162 separately. The inclusion of both the brand inspection and market data from
163 MT provided more complete information on cattle shipments in that state.
164 The datasets are summarized in Table 1 and additional details about the
165 data sets can be found in Appendix B.

166 For each brand inspection (or market) data set, the proportion of in-
167 terstate shipments was calculated. Similarly, the proportion of intracounty
168 shipments (shipments that remain in the county of origin), was calculated.
169 The total number of shipments and the proportion of those shipments that

170 were interstate shipments were separated out by month to examine patterns
171 in seasonality. Because the brand inspection data provided information on
172 the origin and destination locations, we could explore the differences in ge-
173 ographic shipment patterns at the county scale. For each state, the total
174 number of shipments leaving a county was found and the proportion of those
175 that travelled interstate was estimated. Each year of brand inspection data
176 from MT was analyzed separately and the between year correlations were
177 estimated.

178 To examine the relationship between county characteristics and the odds
179 of a shipment traveling to interstate locations, we conducted two logistic re-
180 gression analyses with the odds of shipping to interstate destinations quan-
181 tified in the three brand inspection data sets, and in the MT market data.
182 In these analyses, we considered the total number of shipments leaving a
183 county as a covariate and if the county is located on the state border. In
184 addition to these county characteristics, we examined four measures of the
185 cattle industry in our analyses; these include: the number of operations with
186 cattle inventory, including calves; the inventory of cattle, including calves;
187 the proportion of operations that are beef operations; and the number of feed-
188 lots (operations with cattle on feed). These measures are publicly available
189 through the National Agricultural Statistics Service (NASS) 2012 Census of
190 Agriculture (USDA, 2014) and have been used to inform cattle shipment
191 models (Lindström et al., 2013; Schumm et al., 2015). For each model, we
192 conducted model selection using backwards elimination based on Akaike in-
193 formation criteria (AIC). The full model included the four measures of the
194 cattle industry defined above, an indicator variable for whether the county

195 is a border county, and a variable defining the total number of shipments
196 leaving the county. The final model was selected when no additional terms
197 could be dropped. All continuous covariates were standardized to allow com-
198 parison among predictor variables (Schielzeth, 2010). All models were fit in
199 R version 3.1.0 (R Core Team, 2014).

200 *Comparison of Survey Estimates with Brand Inspection Data*

201 To compare the brand inspection data with the Western region survey
202 results, the brand inspection data had to be combined into a regional es-
203 timate. The brand inspection guidelines for each of the three states differ,
204 however, each data set included information on origin, and destination of the
205 shipments. To make the regional brand inspection data estimates compara-
206 ble with the expert opinion result, we took the mean proportion of interstate
207 shipments across the three states. Additionally, because we had two different
208 intrastate data sets from MT we created two regional brand inspection data
209 sets; the first includes brand inspection data from CA, MT (2010), and WY
210 and the second includes the brand inspection data from CA, WY and the
211 market data from MT (2013).

212 We also compared the expert estimates for each commodity with the
213 brand inspection regional estimates. The state brand inspection records
214 were separated into beef and dairy first and then combined into regional
215 commodity-specific estimates. The beef and dairy designation was already
216 present in the CA brand data so this data set did not require further devel-
217 opment. The brand data from WY included information on the cattle breed.
218 Records for mixed breeds, unknown or unassigned breeds were removed, then
219 the shipments were designated as beef or dairy depending on the breed. The

220 MT brand data did not include a beef or dairy designation and did not pro-
221 vide any information on breed. Therefore, there was no reliable method to
222 separate out dairy shipments; however, we made the assumption that ship-
223 ments of steers were beef shipments and were able to separate those out of
224 the data set (Buhnerkempe et al., 2013). The market data from MT did
225 provide information on cattle breed, so it was possible to assign both beef
226 and dairy designations for this data set in the same way as was done for WY.

227 *Application of Expert Survey Estimates to USAMM Predictions about Tar-*
228 *geted Surveillance*

229 To evaluate the importance of accurately estimating the proportion of
230 inter- versus intrastate shipments, we explored how these proportions im-
231 pacted USAMM model predictions about surveillance and connectivity. The
232 shipment network predictions from USAMM can be used to inform targeted
233 risk-based surveillance of cattle in the U.S. (Gorsich et al., 2018). One group
234 of animals that could be targeted for surveillance are cattle that have been
235 imported from other countries. The importation of live animals is an im-
236 portant route by which diseases could be introduced into the U.S. (Humblet
237 et al., 2009; Tsao et al., 2014). However, these animals are not tracked sep-
238 arately from the rest of the U.S. herd. Imported cattle are given a blue
239 ear tag upon entry, so that they can be easily identified, but these tags can
240 be lost. Gorsich et al. (2018) used the USAMM network to predict where
241 cattle imported from Mexico may be shipped. Here we explored how these
242 predictions may be altered by changing the proportion of shipments that are
243 predicted to travel to interstate versus intrastate locations. We focused on
244 shipments leaving Texas in this study because it was previously identified to

245 have the most counties that receive imports of live cattle from Mexico (34
246 import counties in total) according to the Veterinary Services Import Track-
247 ing system (2009) and Veterinary Services Process Streamlining (VSPS) data
248 (2011) (Gorsich et al., 2018) and because the destination location of ship-
249 ments leaving Texas varied across years (Gorsich et al., 2016). Texas also
250 had the largest number of participants in our expert elicitation survey.

251 We conducted our analyses in three steps. First, we generated the mean
252 USAMM network for shipments originating in TX to use as a baseline for
253 comparison. The mean network was created from 1000 USAMM realizations,
254 each one a simulation of all annual cattle shipments. USAMM predicted the
255 probability of shipments occurring between counties, both within the same
256 state and between counties in different states. The USAMM networks were
257 designed such that the counties were nodes and the shipments between coun-
258 ties were edges (Lindström et al., 2013). The probability that a shipment
259 moved between counties in different states and the number of interstate ship-
260 ments, or edges, predicted by USAMM are informed by ICVI and National
261 Agricultural Statistics Service data. The intrastate shipments predicted by
262 USAMM were estimated by the distance kernel, and therefore, have more un-
263 certainty than the interstate shipment predictions (Lindström et al., 2013).

264 The second step in our analyses was to alter the network such that the
265 proportion of interstate shipments was more in line with the expert mean,
266 minimum and maximum estimates. To do this, we altered the number of
267 intrastate shipments, or intrastate edges, in the predicted mean TX network
268 while holding the number of interstate shipments constant and consistent
269 with the data that informed the model. Changing the network in this way

270 meant that the overall total number of edges in the network changes but the
271 total number of interstate edges did not. The original USAMM TX network
272 predicted that the proportion of interstate shipments was 0.18 (Lindström
273 et al., 2013). The mean expert estimate from TX predicted that the propor-
274 tion of interstate shipments was 0.155, with the range of the expert estimates
275 going from 0.0086 to 0.256. To alter the mean USAMM network, we multi-
276 plied the intrastate shipments by scalars that increased or decreased the total
277 number of intrastate edges, such that the resulting networks had interstate
278 proportions in line with the expert estimates. This preserved the predicted
279 county to county connections, both within and outside of TX and kept the
280 number of interstate edges constant; only the predicted number of intrastate
281 edges, or shipments, in TX changed. We did this for the mean expert es-
282 timates and for the minimum and maximum, which gave us three modified
283 USAMM networks with proportions of interstate shipments of 0.155, 0.0086,
284 and 0.256, respectively.

285 For the third step in our analyses, we used the methods described in Gor-
286 sich et al. (2018), and simulated cattle shipments from the counties receiving
287 imported cattle from Mexico in TX using the original USAMM network and
288 the three modified networks using the expert elicitation data. For these
289 simulations we assumed the probability each imported animal was shipped
290 out of the county that received the imported animals was 1 and varied the
291 probability of not observing an animal, because of random loss of the blue
292 ear tag marking it as an import, from 0 to 1 (Gorsich et al., 2018). We
293 then explored how the three modified networks altered the predicted distri-
294 bution of counties that subsequently receive shipments of imported cattle

295 and if the percent of cattle that could be unobserved while still capturing
296 that distribution changed between the networks. For consistency with the
297 previous methodology and results, we report the same network summary
298 statistics used previously (Gorsich et al., 2018); these include: the number of
299 unique counties reached, the percentages of re-observed cattle in the 10 and
300 50 counties that receive the most shipments, respectively, and the percent of
301 observed cattle moving out of TX, and the skewness and the kurtosis of the
302 distribution of observed cattle among counties receiving shipments.

303 **Results and Discussion**

304 *Expert Elicitation Survey*

305 In total, 51 experts from 19 states and territories participated in the
306 survey (Table B1). The median response rate from the ten focal states (in-
307 cluding experts who were invited and those who responded to the general
308 announcement) was 0.29 (range: 0.1-0.5) and the median final group size
309 from the focal states was 2.5 (range: 1-8) (Table B1). In total, we had
310 seven states with expert group sizes of three or more; these states were Iowa,
311 Minnesota, New York, Oklahoma, Tennessee, Texas and Wisconsin. The re-
312 maining 16 states and territory that responded to the survey announcement
313 had one or two expert participants, which was a limitation of this study. The
314 small number of expert groups with size three or more, was one reason the
315 results were collapsed into regional groups. The regional groups leveraged
316 estimates from multiple state groups and provided more power than the in-
317 dividual state groups, particularly for those states with small sample sizes.
318 The Western region in particular, did not have an individual expert group

319 larger than two; however, the regional estimate included 4 state-level esti-
320 mates. The coverage of expert groups size three or more was better in the
321 Plains, Upper Midwest, Northeast and Southeast. Despite the small number
322 of expert groups size three or more, this study represents the most extensive
323 information on intrastate cattle shipment data in the U.S.

324 Estimates of interstate shipment numbers differed substantially between
325 states and regions in the country. Over all shipment questions, the experts
326 in the plains and northeastern regions estimated the lowest proportion of
327 interstate shipments and the west and southeastern regions were the highest
328 (Figure 1a, Table B2). The national and regional level results for each survey
329 question are presented in the appendix (Tables A1 & A2). The range of es-
330 timates for many questions was large, particularly at the national level. The
331 large variation at the national level was likely due to differences in local and
332 regional shipment patterns, some of which were apparent in the differences
333 between regional estimates from this survey. An additional factor, particu-
334 larly in regions with few survey participants, may have been the low sample
335 size of experts (Table B1). The high variation in question response at the
336 national scale suggests that a single nationwide interstate shipment estimate
337 may not be appropriate and that regional or state-level estimates will be
338 more accurate.

339 The proportion of interstate shipments was calculated for specific ship-
340 ment categories, including market shipments and shipments in the feeding
341 or breeding channel. The estimated proportion of interstate market ship-
342 ments also varied between regions. Experts in the national, upper midwest
343 and northeast regions all estimated that the proportion of market shipments

344 that cross state lines was between 0.36 and 0.45 (Table B2). The experts
345 in the plains region estimated the proportion of interstate shipments was
346 slightly lower at 0.3. Experts in the western and southeast regions both es-
347 timated higher proportions, 0.56 and 0.6, respectively, of interstate market
348 shipments. The feeding channel interstate shipment patterns were estimated
349 to be slightly higher, except for the plains and northeast regions, than for
350 market shipments. However, the general pattern of regional shipment lev-
351 els remained the same. The regions also held similar positions for breeding
352 channel shipments. In general the proportion of interstate breeding channel
353 shipments was lower than both market and feeding channel estimates (Table
354 B2).

355 The survey results from the commodity specific (beef or dairy) sections
356 of the Cattle Movement Survey, also showed geographic variation in the es-
357 timated proportion of interstate cattle shipments. Experts in neighboring
358 states generally estimated similar levels of interstate shipments for beef ship-
359 ments (Figure B.1). The survey results for beef shipments showed regional
360 variation that tended to follow the regional pattern of the estimated overall
361 interstate proportion (Figures 1a-1b, Table B2). Regional patterns of esti-
362 mated proportion of interstate dairy shipments were less well defined than
363 those observed in the interstate beef shipment results and differed slightly
364 in overall pattern from the beef and overall shipment estimates (Figures 1 &
365 B.2, Table B2).

366 As with the interstate overall shipments, we calculated the estimated
367 proportions of specific types of interstate shipments for the commodity spe-
368 cific shipments. The estimated proportions for market, feed and breeding

369 channel commodity specific shipments differed between regions and between
370 commodity type; however, the differences between commodities were not as
371 marked as those between regions (Table B2). Additional descriptions of the
372 commodity specific results can be found in Appendix B.

373 *Brand Inspection*

374 The brand inspection data from CA estimated a proportion of interstate
375 shipments slightly above 50% (Table B3). The brand data from both MT and
376 WY estimated the proportion of interstate shipments lower than CA with all
377 three years falling slightly below 50%. Similarly, the estimated proportion of
378 interstate shipments from the MT market data also showed proportions of
379 interstate shipments slightly below 50%. The multiple years of data from MT
380 showed that the proportion of interstate shipments in the brand data were
381 fairly consistent from year to year and between data sets. Multiple years
382 of data were not available for CA and WY, so they could not be compared
383 through time.

384 The data from the state of Montana were available for a three year period
385 of time (2009-2011). The patterns in the number of shipments originating
386 in each county were very stable across all three years (correlations between
387 years 2009 & 2010: 0.989, 2010 & 2011: 0.982, 2009 & 2011: 0.979). A similar
388 pattern was observed for both number of shipments destined for each county
389 (correlations between years 2009 & 2010: 0.992 , 2010 & 2011: 0.990, 2009
390 & 2011: 0.986) and for the proportions of interstate shipments (correlations
391 between years 2009 & 2010: 0.894, 2010 & 2011: 0.962, 2009 & 2011: 0.854).

392 The total number of shipments per month showed bimodal seasonality,
393 with peaks in the spring (April to May) and in the fall (Oct. to Nov.) for all

394 years of MT brand, MT market and WY brand data (Figure 2a). The CA
395 brand data showed a similar spring peak in total number of shipments but
396 did not have a second peak in the fall. The proportion of interstate shipments
397 did not scale directly with the total number of shipments for MT or WY, and
398 therefore showed a different pattern in seasonality in these states (Figure 2b).
399 For the MT data sets (both brand and market), and the WY brand data,
400 the proportion of interstate shipments had a single peak in the fall months
401 (Sept. to Nov.). This was particularly apparent in the MT brand data which
402 reported the lowest proportion of interstate shipments in the spring and the
403 highest in the fall. The proportion of interstate shipments reported in the
404 CA brand data did not follow the same pattern as the other states. In CA,
405 the proportion of interstate shipments peaked in May and corresponded with
406 the peak in the total number of shipments.

407 For all three states, the brand inspection data showed that there were
408 differences in the number of outgoing shipments between counties within the
409 respective states (Figures 3a, 3c, 3e, & B.3a, B.3c). The proportion of inter-
410 state shipments also varied between counties in the same state (Figures 3b,
411 3d, 3f, & B.3b, B.3d). The odds of counties shipping to interstate destinations
412 in the brand inspection data were influenced by all covariates considered, but
413 the magnitude and direction of each co-variate varied by state (Figure 4a).
414 In CA, the best predictors were the total number of shipments and the num-
415 ber of feedlots (operations with cattle on feed). In MT and WY, the best
416 predictors were whether the county was on a border, the total number of
417 shipments, and the proportion of operations in the county that were beef.
418 Border counties consistently shipped more out of state shipments, 1.12, 2.17,

419 and 3.03 times higher odds of shipping out of state in CA, MT, and WY,
420 respectively (95% CI CA: 1.03–1.22; MT: 2.07–2.28; WY: 2.70–3.41). In con-
421 trast, associations with the total number of shipments and the proportion of
422 operations with beef cattle were variable by state. In MT, counties sending
423 more shipments and those with a higher proportion of beef operations were
424 more likely to ship interstate while in WY, counties with a higher proportion
425 of beef shipments were less likely to ship interstate.

426 Similar to the brand data, there was variation between market counties
427 in both total outgoing shipments and proportion of interstate shipments in
428 the MT market data (Figure B.4). The final model predicting the odds
429 of shipping to interstate destinations in the MT market data included the
430 number of cattle operations, the proportion of operations that are beef, the
431 total inventory of cattle, and the total number of shipments leaving that
432 county. The best predictors were the total number of shipments, the total
433 inventory of cattle in the county, and the proportion of operations that are
434 beef (Figure 4b). Market counties with one standard deviation more cattle
435 were associated with a 1.45 times higher odds of shipping interstate (95%
436 CI: 1.38–1.52) and counties with higher proportions of beef operations were
437 associated with a 1.27 times higher odds of shipping interstate (95% CI:
438 1.07–1.50). Conversely, counties with markets sending a larger number of
439 shipments were less likely to send out of state, as one standard deviation
440 more shipments was associated with a 0.65 times lower odds of shipping
441 interstate (95% CI: 0.60–0.71).

442 The brand inspection data provided detailed information on within and
443 between state shipments for three western states, CA, MT and WY. The

444 level of detail in the data sets allowed us to investigate both the proportion
445 of interstate shipments and the proportion of intracounty shipments (ship-
446 ments that remain within the county of origin). Additionally, we were able
447 to explore the temporal and geographic differences in the number of outgoing
448 shipments and the proportion of those which were interstate at the monthly
449 and county level, respectively. The temporal patterns in the total number
450 of shipments originating in MT (both brand and market data sets) and WY
451 followed the same bimodal pattern of shipments peaking in spring and fall
452 that was reported in ICVIs (Gorsich et al., 2016). CA showed the same
453 spring peak in shipments but did not show the second fall peak. The differ-
454 ences in these temporal patterns between states could be attributed to the
455 differences in brand inspection requirements (California Department of Food
456 and Agriculture, 2017; Montana.gov Official State Website, 2017; Wyoming
457 Livestock Board, 2017), differences in the cattle industry or a combination of
458 both. These data sets also provided a unique look at the temporal changes
459 in the proportion of interstate shipments in different states. In CA the pro-
460 portion of interstate shipments increased at the same time the total number
461 of shipments increased. However, for MT and WY the seasonal patterns of
462 the proportion of interstate shipments did not follow the total number of
463 shipments. These data suggested that in the fall the proportion of interstate
464 shipments increases. The pattern of seasonality in the proportion of inter-
465 state shipments could affect the potential for cross state border spread of
466 disease outbreaks, such that chance of long distance spread could increase
467 during the seasons when the proportion of interstate shipments peaks.

468 We explored the geographic differences in the total number of shipments

469 and the proportion of interstate shipments at the county level for each brand
470 and market data set. We found that all the covariates that we considered
471 influenced the odds of shipping interstate, but that these covariates acted
472 in different ways and to varying degrees depending on the state. However,
473 our analysis suggested that border counties have higher odds of shipping to
474 interstate destinations in all three brand inspection data sets. In both the
475 brand inspection and market data analysis the total number of shipments and
476 proportion of beef operations were important covariates but they acted on
477 the odds of interstate shipping in different ways. Interestingly, the covariates
478 varied between the MT data sets (brand and market) as well. The cattle
479 inventory seemed to be more important for determining interstate shipment
480 odds in the market data than in the brand data, and total shipments had a
481 positive influence on the brand data and a negative influence on the market
482 data. This could suggest that large beef movements use markets. These
483 results also suggested that the proportion of shipments that leave counties
484 are correlated to the total number of shipments and to other indicators of
485 the cattle industry and infrastructure, such as proportion of operations which
486 are beef.

487 County level heterogeneity was also found in analyses of cattle shipment
488 networks based on ICVI data (Buhnerkempe et al., 2013). Buhnerkempe
489 et al. (2013) found that though the cattle shipment network was highly con-
490 nected, the county level heterogeneity was such that state-level networks
491 would most likely be too coarse for examining disease outbreaks. The brand
492 inspection data sets showed similar patterns in county level heterogeneity
493 some of which was explained by the total number of shipments leaving a

494 county, and additional covariates. This indicated that interstate shipment
495 data, such as ICVIs, in combination with generally available covariates such
496 as, proximity to a border, the proportion of beef operations and potentially
497 the presence of a market or feedlot, can be used to inform intrastate ship-
498 ment predictions such as those developed by Lindström et al. (2013). Given
499 that the overall estimate of the proportion of interstate shipments from the
500 experts in the western region was close to that of the brand data, it is pos-
501 sible that expert estimates, though on a much coarser scale than brand or
502 NASS data, could also be used to help inform shipment patterns in areas of
503 the U.S. where additional intrastate data are unavailable. This has impor-
504 tant implications for development of national-scale cattle shipment models
505 with the objective of modeling disease spread (Buhnerkempe et al., 2014) or
506 for identifying counties and states of increased risk for receiving shipments
507 of at-risk animals (Gorsich et al., 2018). These findings make the develop-
508 ment of national-scale shipment predictions more tractable because within
509 state shipment data are not available for most of the U.S. and ICVI data are
510 currently the best source for all regions of the U.S. This also has potential
511 implications for foreign animal disease preparedness planning in that coun-
512 ties that connect within state shipment patterns to interstate shipments can
513 be identified based on number of interstate shipments and covariates that are
514 easily accessible. This information alone is valuable for planning surveillance
515 activities or risk mitigations such as movement controls when detailed infor-
516 mation is not available or too time consuming to develop during a emergency
517 response event.

518 Finally, the brand data provided an opportunity to explore the possibility

519 of changes to the proportion of interstate shipments through the year. Sur-
520 prisingly, the proportion of interstate shipments did not directly follow the
521 seasonality in total number of shipments for MT or WY. This suggests that
522 there may be differences in interstate shipment seasonality in other states as
523 well; information that could be very valuable in determining the probability
524 of a disease spreading over state lines.

525 *Comparison of Survey Estimates with Brand Inspection Data*

526 The comparison between the western region (CA, ID, MT, NV) expert
527 elicitation survey results and the western region brand inspection results
528 (CA, MT Brand 2010, & WY, and CA, MT Market, & WY, respectively) on
529 the proportion of interstate shipments showed that estimates from these two
530 data sets were quite similar (Figure 5). The similarity between the overall
531 survey estimate and the brand inspection data suggested that the overall
532 estimated level of interstate shipments by region were in the range of the
533 observed number.

534 The brand inspection regional estimates changed slightly when the data
535 were broken out into beef and dairy commodity types, with the propor-
536 tion of beef interstate shipments remaining close to the overall estimate and
537 the dairy estimate increasing. However, the expert elicitation results when
538 broken out into beef and dairy changed more substantially, with the esti-
539 mated proportion of interstate shipments increasing for beef and decreasing
540 for dairy. The brand inspection data and expert survey estimates for the
541 commodity specific (beef or dairy) proportions of interstate shipments did
542 not agree as well as they did for the overall estimate; the estimates for dairy
543 were particularly divergent. This could suggest that the shipment patterns

544 of individual commodities, especially dairy, are less well understood than the
545 overall shipment patterns.

546 The variation within the commodity specific expert estimates, and the
547 comparison of these estimates with brand inspection data suggested that
548 the the amount of interstate shipments between beef and dairy is less well
549 understood, at least for western states, than the overall level of interstate
550 shipments. The high degree of variation in the results of the expert survey
551 may be caused by more than uncertainty in the system; different interpre-
552 tations of the questions and the clarity of the questions being asked could
553 also play a role in the amount of variation seen in the results. Gathering
554 additional commodity specific shipment data will help identify causes of un-
555 certainty and will be beneficial for building data driven shipment models and
556 for developing effective response plans.

557 Expert estimates on proportion of interstate shipment varied regionally
558 in the United States. Similarly, experts estimated that differences exist in
559 the proportion of interstate shipments between the cattle commodities, beef
560 and dairy. Though we were unable to do a comprehensive validation of the
561 expert estimates, we were able to compare the western region to the regional
562 brand inspection data. We found that the mean expert estimate for overall
563 proportion of interstate shipments was similar to the brand inspection esti-
564 mate, but that the commodity specific expert estimates were more divergent
565 from the brand estimate. This large variation for some types of shipments
566 may indicate that certain aspects of the cattle industry are generally less
567 understood or that there is a diversity of mechanisms that influence ship-
568 ments for some parts of the cattle industry and that no one expert possessed

569 all of the information. This large variation also highlights the importance
570 of developing empirical data to inform descriptions of cattle shipments and
571 that relying solely on expert knowledge could provide biased estimates. This
572 could also have implications for other types of livestock shipment models
573 that rely heavily on expert opinion (Pines et al., 2007; Wongsathapornchai
574 et al., 2008).

575 Our results identify several aspects of intrastate shipments in the U.S.
576 that may not be well understood. It is generally thought that different re-
577 gions of the country have different cattle shipment patterns and the empirical
578 interstate data suggest that this is true (Gorsich et al., 2016). The expert
579 estimates support the theory that different regions have different shipment
580 patterns, though due to the small sample size of some states and regions it
581 is difficult to verify the regional pattern with these data. The differences
582 in shipment seasonality that were present in the brand inspection data be-
583 tween states also suggest that there are differences between states that could
584 lead to regional differences in shipment patterns. However, because we only
585 have empirical data for the western U.S., we are unable to fully validate how
586 shipments might vary by region in the U.S. Similarly, the differences in the
587 expert estimated proportions of different types of interstate cattle shipments
588 (i.e. market, feeding channel or breeding channel shipments) are not fully
589 observable in the brand inspection data. Gathering empirical data to sup-
590 port or refute regional and shipment type differences in the proportions of
591 interstate shipments would be a valuable addition for both modeling and
592 decision-making efforts.

593 *Application of Expert Survey Estimates to USAMM Predictions about Tar-*
594 *geted Surveillance*

595 The simulations of imported cattle to TX for the original USAMM net-
596 work and the three modified networks suggested that while the skewness and
597 kurtosis were variable across the differing levels of proportion of interstate
598 shipments, the predicted total number of unique counties reached, and the
599 percent of imported cattle re-observed in the 10 and 50 counties that receive
600 the most shipments, respectively, were fairly stable (Table B4). Similarly,
601 the distribution of unique counties reached were similar and fairly stable
602 until around 90% of the cattle are unobserved (Figure 6) for all four net-
603 works. These patterns are consistent with those reported when using the full
604 USAMM network rather than just a shipment originating in a single state
605 (Gorsich et al., 2018). The modified network with the proportion of interstate
606 shipments corresponding to the expert estimate minimum (0.0086) showed
607 the most difference from the original USAMM network. The predicted num-
608 ber of unique counties reached was substantially lower (on average 41% lower)
609 than the other networks; however, the shape of the distribution was similar
610 to those predicted by the other networks.

611 The other clear difference between the original network predictions from
612 TX and the modified networks was the percent of cattle predicted to leave
613 TX, the state of importation. The prediction from the original TX network
614 was that 50.28% of cattle (individual animals, not shipments) will leave the
615 state when all cattle are observed. The predictions from the modified net-
616 works ranged from 5.73% to 59.01% cattle leaving TX, when all cattle are
617 observed. These predicted percentages were fairly consistent over the varying

618 levels of the percent of observed cattle. While the difference in predictions
619 between the maximum and minimum expert estimates were considerable,
620 there did appear to be some robustness to uncertainty in the proportion
621 of interstate shipments in the system. The predictions from the mean and
622 maximum modified network and the original network ranged from 46.80% to
623 59.01% cattle leaving TX. This would suggest that for some range of TX pro-
624 portions, the predictions of cattle leaving the state would not substantially
625 change. However, the minimum expert estimate cannot be completely dis-
626 missed as an outlier since there is currently no observed intrastate data from
627 TX to compare to and from a surveillance perspective the difference between
628 94% of cattle remaining in their state of importation versus 41% of the cattle
629 remaining is an important difference to examine. The network connections
630 and the distribution of the network summary statistics were fairly consistent
631 between the four explored networks, but the amount of resources (e.g. num-
632 ber of tests, staff) required for surveillance and the spatial distribution of
633 those resources could be altered depending on the proportion of interstate
634 shipments. More importantly the change in the expected number of cattle
635 remaining in their state of importation could significantly alter surveillance
636 strategies and interpretation of surveillance results because sample sizes re-
637 quired may be based on the wrong number of animals.

638 The data sets and results we present here indicate the importance of
639 understanding intra- and interstate shipment patterns. The relationship be-
640 tween intra- and interstate shipment patterns we observed and their con-
641 sistency with previous analyses of national-scale shipment patterns (Buhn-
642 erkempe et al., 2013) provides evidence that current methods to predict cattle

643 shipments such as those developed by Lindström et al. (2013) and the appli-
644 cation of these methods to predict movement of at-risk animals are consistent
645 with industry shipment patterns. Additionally, our results can be used to
646 identify aspects of cattle shipment practices that require additional study
647 and data collection, such as the characterization of regional-, temporal-, and
648 commodity-specific shipment patterns.

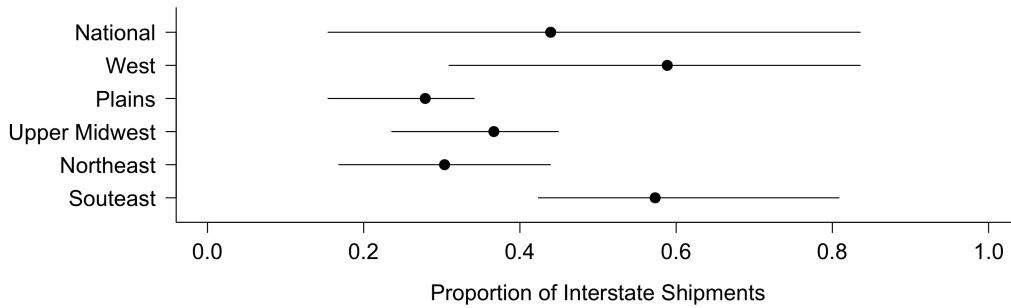
649 **Conclusions**

650 The development of and comparisons among these four data sets is an im-
651 portant step for improving our understanding of intrastate cattle shipments
652 in the United States. Our results both corroborate existing literature that
653 predicts U.S. cattle shipments and indicate that regional differences exist
654 in cattle shipments as well as highlight potential gaps in current knowledge
655 about cattle shipment patterns and industry practices. As we demonstrate
656 with our application of expert data to targeted surveillance of import cattle
657 in TX, the data sets developed here can also be used to inform modeling
658 efforts, such as the previously developed models on cattle shipments and
659 disease spread (USAMM and USDOS), which can be used for national-level
660 preparedness and response plans, as well as for tracing and surveillance ap-
661 plications.

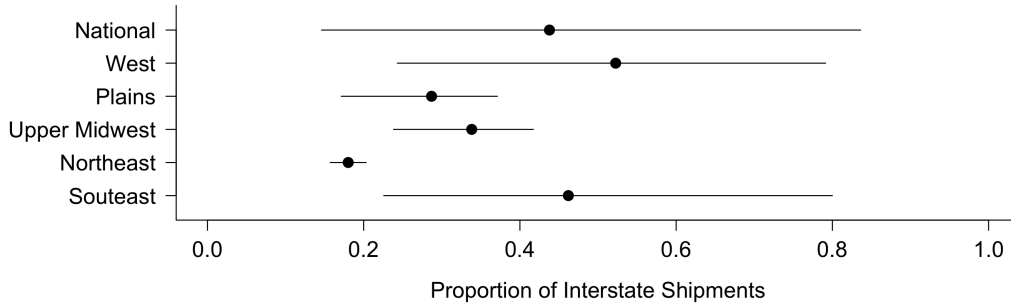
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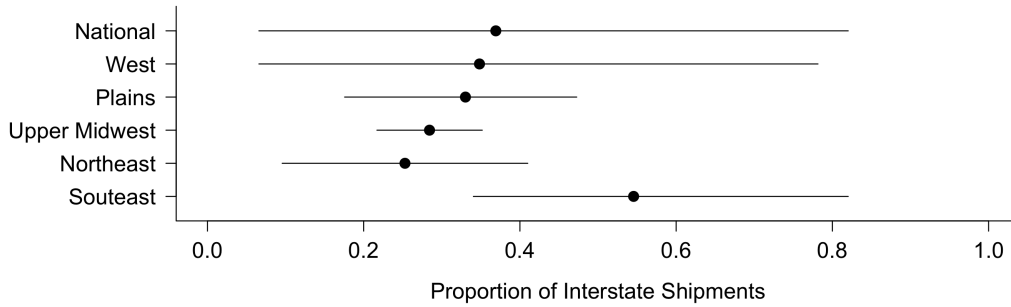
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677 contained in this document are those of the authors and should not be in-
678 terpreted as necessarily representing the regulatory opinions, official policies,
679 either expressed or implied, of the USDA-APHIS-Veterinary Services or the
680 U.S. Department of Homeland Security.



(a) Proportion of Interstate Shipments

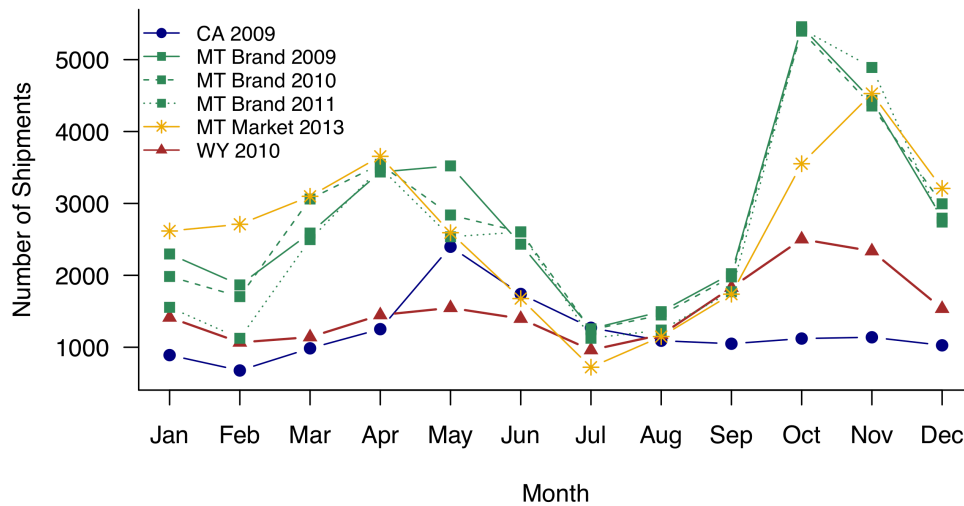


(b) Proportion of Interstate Beef Shipments

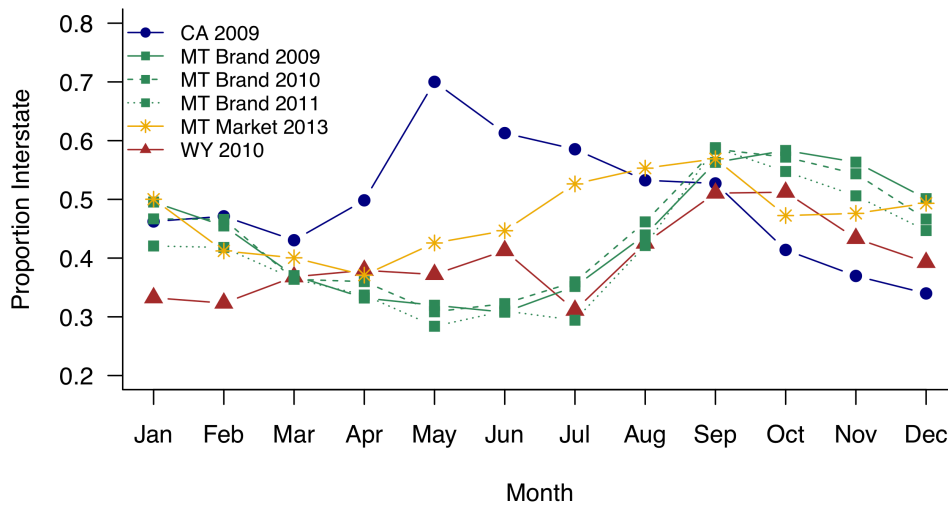


(c) Proportion of Interstate Dairy Shipments

Figure 1: **Proportion of interstate shipments by region.** a) The proportion of interstate shipment overall by region. b) The proportion of interstate beef shipments and c) the proportion of interstate dairy shipments by region. The ordering of the national and regional estimates is the same for all three plots. The black point shows the mean of all cattle (a), or beef cattle (b) or dairy cattle (c). The lines show the range of expert estimates.



(a) Total Shipments by Month



(b) Proportion of Interstate Shipments by Month

Figure 2: **Shipment characteristics by month.** a) The number of total outgoing shipments (intra- and interstate) by month. b) The proportion of shipments that travel to interstate destinations by month. The different points and colored lines represent the four different brand inspection and market data sets. The different years in the MT brand data are shown with different types of lines. CA: navy, circles; MT brand: green, squares; MT market: yellow, stars; WY: red, triangles.

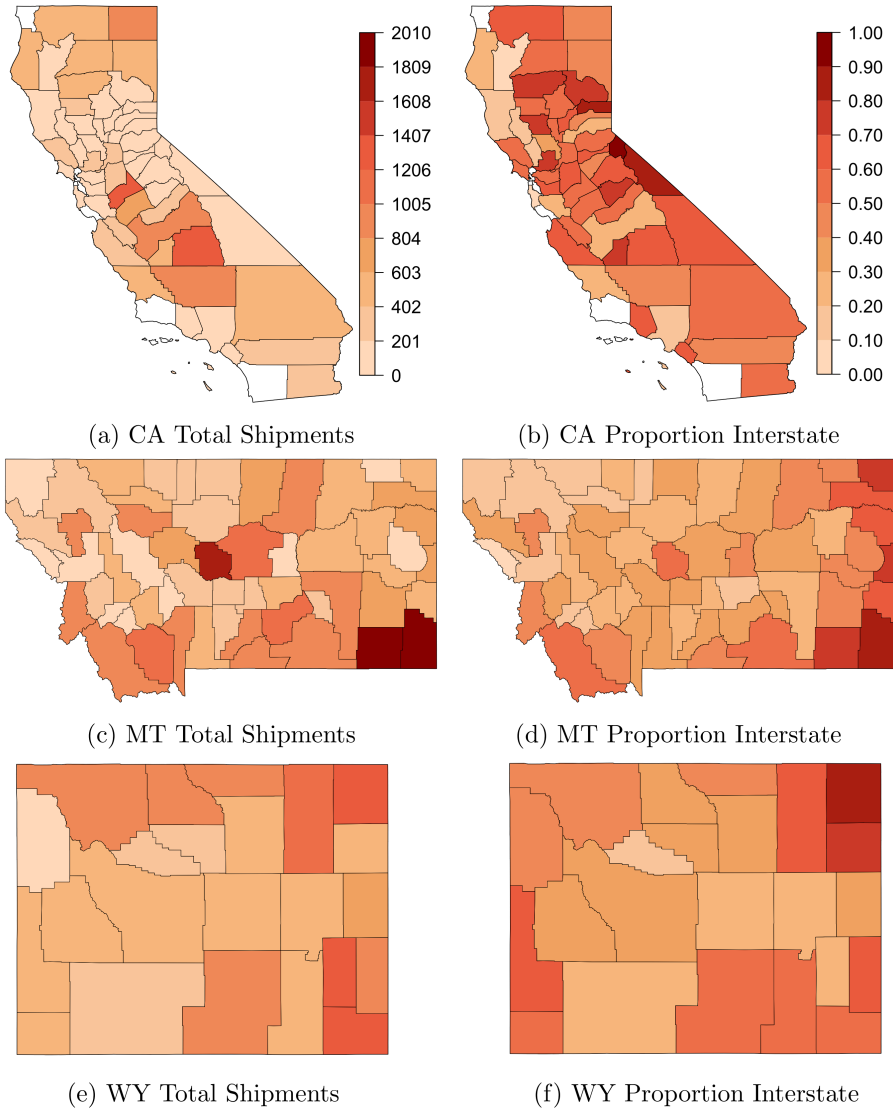


Figure 3: **Brand inspection county shipment characteristics by state.** a, c, e) The number of total outgoing shipments (intra- and interstate) by county. b, d, f) The proportion of interstate shipments by county. The scale increases moving from light orange to dark red. Note that the scale of the legend changes between the to total shipment and proportion of interstate shipment plots. Counties shaded in white have no data. Panels a & b show CA, c & d show MT 2010 and e & f show WY.

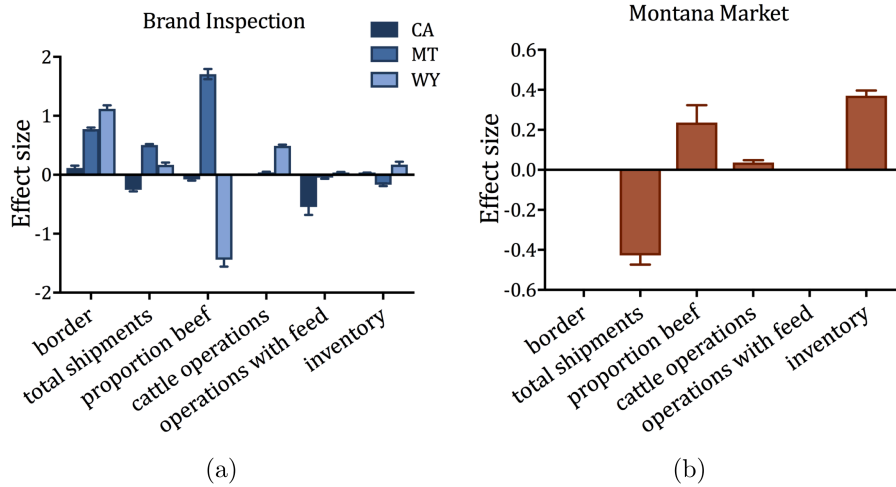
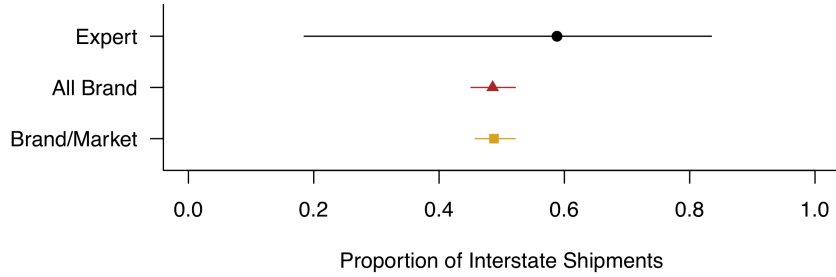
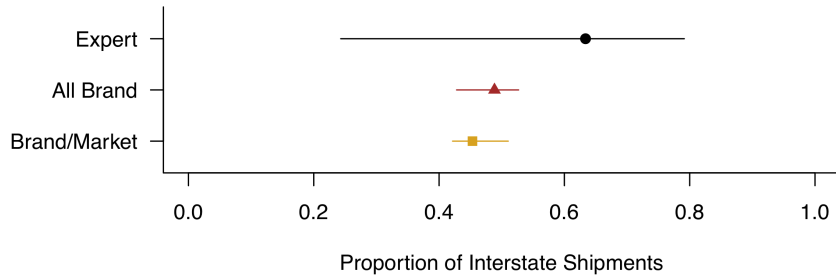


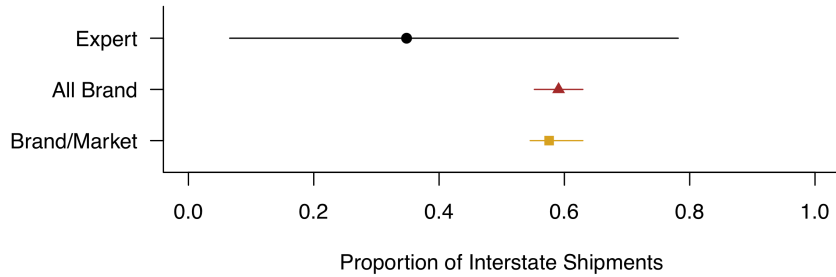
Figure 4: **Effect size and standard error for standardized co-variables in logistic regression analyses.** The analyses predict the odds of out-of-state shipment in a) the brand inspection data sets (CA, MT 2010 & WY) and b) the MT market dataset. Independent variables are displayed on the x-axis and represent an indicator variable for whether the county is on the state border (border), the total number of shipments in the dataset (total shipments), the proportion of operations that were beef (proportion beef), the number of operations on feed, the total number of operations with cattle, including calves (cattle operations), and the total inventory of cattle (inventory). All coefficients retained in the model were significant. Note that the y-axes on the two plots are on different scales.



(a) Proportion of Interstate Shipments



(b) Proportion of Interstate Beef Shipments



(c) Proportion of Interstate Dairy Shipments

Figure 5: **Proportion of interstate shipments in the West.** a) The proportion of interstate shipments overall. b) The proportion of interstate beef shipments and c) the proportion of interstate dairy shipments. The black points are Western region (CA, ID, MT, and NV) expert elicitation survey data, the dark red points are regional brand inspection (CA, MT and WY) results and the orange points are regional brand inspection (CA and WY) and market data (MT) results. The lines show the ranges of the expert estimates and the state-level brand inspection results, respectively.

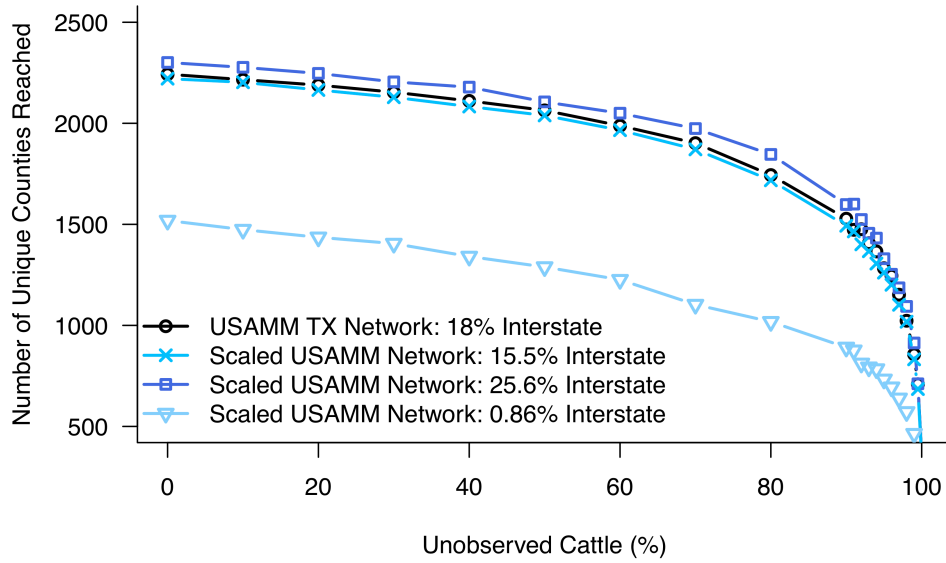


Figure 6: **The predicted number of unique counties reached after shipment from initial import county.** The lines show the predictions by the original TX USAMM network (black line, circles), and the TX network with interstate proportions scaled to be in line with the expert survey predicted mean (teal line, X's) and range (min: light blue line, triangles; max: dark blue line, squares).

Table 1: Summary of the Brand Inspection and Market Data.

State	Year	Reasons for Inspection	Information in Data	Reference
California (CA)	2009	Change of ownership; Interstate, slaughter, or market shipments; Entering feedlots; Movements out of specific designated areas	Inter- or intrastate shipment; commodity type	Bureau of Livestock Identification, (2017)
Montana (MT)	2009 to 2011	Change of ownership; Inter-county or state shipments; prior to slaughter or auction	Inter- or intrastate shipment; reason for the movement	MT Department of Livestock (2017)
MT market data	2013	Animal-level records of market shipments	Inter- or intrastate shipment; name of market; breed	MT Department of Livestock (2017)
Wyoming (WY)	2010	Change of ownership; Inter-county or state shipments; shipments to markets	Inter- or intrastate shipment; breed; purpose of shipment	WY Livestock Board (2017)

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