

1 **Long-term surveillance defines spatial and temporal patterns**
2 **implicating *Culex tarsalis* as the primary vector of West Nile virus in**
3 **Iowa, USA**

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16 Running Title: Patterns of WNV transmission in Iowa

17 **Abstract**

18 West Nile virus (WNV) has become the most epidemiologically important mosquito-
19 borne disease in the United States, causing ~50,000 cases since its introduction in
20 1999. Transmitted primarily by *Culex* species, WNV transmission requires the complex
21 interplay between bird reservoirs and mosquito vectors, with human cases the result of
22 epizootic spillover. To better understand the intrinsic factors that drive these
23 interactions, we have compiled infection data from sentinel chickens, mosquito vectors,
24 and human cases in Iowa over a 15 year period (2002-2016) to better understand the
25 spatial and temporal components that drive WNV transmission. Supplementing these
26 findings with mosquito abundance, distribution, and host preferences data, we provide
27 strong support that *Culex tarsalis* is the most important vector of human WNV infections
28 in the region. Together, our analysis provides new insights into WNV infection patterns
29 in multiple hosts and highlights the importance of long-term surveillance to understand
30 the dynamics of mosquito-borne-disease transmission.

31 **Introduction**

32 In an era of increased concern over mosquito-borne viruses, West Nile virus (WNV)
33 continues to have the largest epidemiological impact in the United States, causing
34 ~50,000 cases and over 2,100 deaths since its introduction in 1999 (1). Therefore, the
35 annual occurrence of WNV presents a continual public health threat across the
36 continent.

37 WNV persists in nature through a host cycle that involves bird reservoirs and *Culex*
38 mosquito vectors, with humans serving as dead-end hosts through the bite of an
39 infected mosquito (2). Both abiotic factors (e.g.- landscape, climate, and seasonality)
40 and biotic factors (e.g.- community composition of bird reservoirs and mosquito vectors,
41 species abundance, and inter-species interactions) influence WNV infection dynamics
42 across the United States (3–7). Human cases of WNV have been reported across all 48
43 states of the contiguous US, yet specific states or geographic regions have been
44 impacted by a disproportionate number of cases (1). Three states (California, Colorado,
45 and Texas) comprise more than 36% of the total number of reported WNV cases, while
46 states of the Upper Midwest, particularly South Dakota and Nebraska, have consistently
47 displayed among the highest incidence rates (cases per capita) in the country (1). As a
48 result, it is of great importance to understand the epidemiological patterns of WNV
49 across the country (8).

50 Previous reports describing patterns of WNV transmission in densely populated
51 urban/suburban areas (9–12) and rural environments (13, 14), suggest that distinct
52 mechanisms of peridomestic and sylvatic transmission may influence the epidemiology
53 of WNV across the US (8). This is supported by considerations of land use and
54 landscape ecology, which serve as important determinants in shaping the geographical
55 distributions of mosquito vectors (15–19). Of the *Culex* species implicated as vectors of
56 WNV, *Culex pipiens*, *Culex quinquefasciatus*, and *Culex restuans* are routinely
57 associated with urbanized settings (20, 21) while *Culex tarsalis* is most common in rural
58 habitats (15, 16, 22). With known differences in vector competence (23, 24), the spatial
59 distributions of these *Culex* species can have profound impacts on patterns of WNV
60 transmission at the local or regional level. Factors influencing WNV transmission have

61 been broadly described across the United States (8, 25–27), yet our understanding of
62 WNV transmission in the Midwest has heavily relied on studies of the Chicago
63 metropolitan area (12, 28–33) with only limited characterization of WNV epidemiology
64 and transmission dynamics in other locations of the Midwest (14, 15, 34–36).

65 Building on initial reports describing WNV transmission in Iowa from 2002-2006 (15), we
66 have assembled an extensive 15 year study of WNV transmission for the state (2002 -
67 2016) examining WNV infection data from human, sentinel chicken, and mosquito
68 hosts. Combined with mosquito abundance, distribution, and host selection data, our
69 comprehensive resources offer the unique ability to investigate the concurrent impacts
70 of multiple ecological factors on human WNV disease cases. By examining host
71 infection rates from spatial and temporal perspectives, we have determined when and
72 where WNV is most actively being transmitted in Iowa, which suggests underlying biotic
73 and abiotic mechanisms that likely influence WNV transmission. Together, these data
74 provide strong support that *Cx. tarsalis* is the predominant vector of human WNV
75 transmission in Iowa. These analyses provide new insights into WNV transmission
76 dynamics in the Upper Midwest region and highlight the importance of long-term
77 surveillance to understand mosquito borne-disease.

78 **Materials & Methods**

79 **West Nile virus surveillance program**

80 Since 2002, the Medical Entomology Lab at Iowa State University has conducted WNV
81 surveillance for the state of Iowa through cooperative efforts with the Iowa Department
82 of Public Health, the State Hygienic Lab, and city- or county-level municipalities around
83 the state, where field operations occurred. Data presented herein were collected over
84 the 15 year period from 2002 to 2016, generally between mid-May and early October
85 (approximately weeks 20 - 42) of each year as the period of mosquito activity in Iowa.

86 **Mosquito collections and identification**

87 Adult mosquitoes were collected across the state using a variety of trapping methods to
88 assess population dynamics or to monitor WNV infection rates. New Jersey light traps
89 (NJLTs) were used to measure mosquito abundance, while CO₂-baited CDC light traps

90 (CDC traps) or grass infusion-baited gravid traps were used to collect mosquitoes for
91 subsequent WNV testing. All three trap types were used for the duration of the 15 year
92 study period, with the exception of a two year period in which CDC trapping did not
93 occur (2014 and 2015) and one year (2002) in which gravid traps were not used.
94 Additional trapping methods relying on Mosquito Magnets (2007–2009) or BG-Sentinel
95 traps (2016) were selectively used across the state. All traps were maintained and
96 operated by either Iowa State University personnel or local municipal collaborators.

97 Mosquito samples were identified according to morphological characteristics described
98 for North American mosquitoes (37). To keep data consistent across the 15 year
99 trapping period and the difficulty in distinguishing adult female *Cx. pipiens*, *Cx.*
100 *quinquefasciatus*, *Cx. restuans*, and *Cx. salinarius* (especially when samples are in poor
101 condition), these species were collectively identified as “*Cx. pipiens* group” due to the
102 difficulty in distinguishing these species as adults, especially in the large volumes
103 processed in our study with specimens frequently in poor physical condition (38, 39).
104 This nomenclature was used throughout the study.

105 **West Nile virus (WNV) testing in mosquitoes**

106 Mosquito samples collected from CDC and gravid traps were routinely used for WNV
107 testing during the 15 year sample period. Additional samples collected from BG-Sentinel
108 and Mosquito Magnet traps were respectively tested for only a single season or a small
109 period of 3 years. Following collection, mosquitoes were stored at either -20 °C or -80
110 °C by our collaborators before shipping in insulated parcels with frozen ice packs or dry
111 ice. To preserve samples during identification and handling, mosquitoes were sorted on
112 refrigerated tabletops. After identification, mosquitoes of the same species were
113 grouped together into pools not exceeding 50 mosquitoes according to trap type, trap
114 site and collection week. Mosquito pools were sent to the State Hygienic Laboratory
115 (Iowa City, IA) for WNV detection by quantitative RT-PCR. Only data collected for *Culex*
116 species that regularly tested positive for WNV (i.e., *Cx. pipiens* group, *Cx. tarsalis*, and
117 *Cx. erraticus*) were used in this study and were collectively referred to as “*Culex*”
118 species (Table S1).

119 Mosquito infection rates were calculated as minimum infection rates (MIR) from bias-
120 corrected maximum likelihood estimates using the CDC-provided Microsoft Excel add-in
121 for calculation of pooled infection rates (40). Mosquito infection data were parsed
122 according to week of the year, mosquito species, trap type, and county of collection.

123 **Sentinel chicken seroprevalence**

124 Sentinel chickens were used in participating Iowa municipalities from 2002 to 2013 to
125 monitor WNV seroprevalence in local wild bird reservoirs. At the beginning of each
126 surveillance year, baseline bleeds were performed on each chicken to ensure the
127 absence of WNV infection. Sentinel “flocks” consisting of 8 chickens were deployed to
128 sites around the state, based on the availability of municipal collaborators to maintain
129 the flocks. Weekly blood draws from each individual chicken (target volume – 0.5 mL)
130 were taken to gauge infection status through the summer. All blood samples were sent
131 to the State Hygienic Laboratory (Iowa City, IA) to test for the detection of IgM
132 antibodies to WNV using a MAC-ELISA assay as previously described (41).
133 Seroprevalence was calculated by dividing the number of infected chickens by the total
134 number of sampled chickens in a population. Animal procedures were approved by the
135 Iowa State University Institutional Animal Care and Use Committee protocol #7-2-
136 5196G to L. Bartholomay.

137 **Human case reporting**

138 As a mandatory reportable disease, human cases of WNV were reported to the Iowa
139 Department of Public Health from physicians and health clinics around the state, before
140 forwarding information to the Centers for Disease Control (CDC). All human case data
141 were reported at the county level by week of the year (except 2002, which lacked
142 temporal data). Human incidence was calculated by dividing the number of reported
143 human cases by the human population (Iowa county or region) and normalized per
144 100,000 people.

145 **Mosquito blood meal analysis**

146 Mosquitoes containing a visible blood meal that were collected in CDC light traps or
147 gravid traps at various locations throughout the state were preserved at -80°C. This

148 included all *Cx. tarsalis* blood-fed samples in our possession collected from 2007 to
149 2011 and in 2017, a total of 93 samples. A selection of 214 blood-fed *Cx. pipiens* group
150 samples collected from 2015 and 2017 were similarly processed. DNA was isolated
151 from individual mosquitoes using the Marriot DNA extraction procedure (42). DNA from
152 the *Cx. tarsalis* samples were validated using actin primers (43), while *Cx. pipiens* group
153 samples were speciated by PCR analyses as previously (44) to confirm their
154 identification as *Culex restuans*, *Culex pipiens*, or *Culex salinarius*. Positive DNA
155 amplification was confirmed in 92 of 93 *Cx. tarsalis* and 159 of 214 *Cx. pipiens* group
156 samples.

157 To determine the vertebrate host source of the mosquito blood meals, two methods
158 were employed. For *Cx. tarsalis*, a PCR-RFLP assay in which cytochrome B (cytB)
159 sequences were amplified by PCR, then digested with *TaqI* to distinguish between
160 vertebrate species was performed on the samples (45). However, due to similarities in
161 the *Culex pipiens* group cytB sequences with the associated PCR-RFLP primers, this
162 assay could not distinguish vertebrate blood samples. As a result, the *Culex pipiens*
163 group samples were evaluated by PCR using individual avian, mammal, and human
164 primer sets as previously (31, 46). PCR reactions were performed using DreamTaq
165 (Thermo Fisher) and visualized by electrophoresis on 1.5% agarose gels.

166 **Mapping of WNV infection data and mosquito abundance data**

167 To visualize host (human, chicken, and mosquito) infection rates and mosquito
168 abundance on a geographic scale, county-specific values were mapped using ArcMap
169 10.4.1 (Esri ArcGIS). Values of human incidence, sentinel chicken seroprevalence,
170 mosquito MIR, and mosquito abundance ratios were interpolated between counties
171 according to the inverse distance weighted (IDW) method using centroid locations of the
172 corresponding counties as previously described (47). To enhance the visual display, a
173 2% clip and gamma stretch were applied.

174 **Statistical analyses**

175 All variables examined in this study (host infection rates by week, year, region, and
176 county; county-level mosquito abundance measures) were found to have non-normal

177 distributions, with the exception of regional mosquito MIR differences. As a result, non-
178 parametric methods were chosen for all statistical analyses to maintain consistency.
179 Significance of relationships was determined with a 95% confidence interval ($P < 0.05$,
180 two-tailed). A Wilcoxon signed-rank test was performed to determine differences in
181 annual host infection rates between Iowa regions and between *Culex* species. Data
182 were analyzed and prepared using GraphPad Prism 6 (GraphPad Software).

183 Results

184 WNV displays defined temporal patterns in its human, avian, and mosquito hosts

185 To better understand the seasonality of WNV transmission in Iowa, fifteen years of
186 WNV surveillance data were analyzed from the three major host species (humans,
187 birds, and mosquitoes) to identify peak periods of WNV activity. Between mid-May and
188 early October (approximately weeks 20-42), mean weekly WNV activity is displayed for
189 human cases (Fig. 1A), chicken seroprevalence (Fig.1B), and mosquito infection rates
190 (Fig. 1C). Across host species, WNV activity followed similar temporal patterns reaching
191 detectable levels approximately in early June (week 23) and amplifying progressively
192 through the summer months. Human and mosquito infection rates peaked in early
193 September (humans, week 36; mosquito, week 37), then rapidly declined to barely
194 detectable levels in October (week 40). In contrast, chicken seroprevalence continued
195 to rise through late September and early October when human incidence and mosquito
196 infection rates have already rapidly declined (Fig. 1). Seroprevalence was not measured
197 beyond this point as it marked the end of the field season.

198 Relationships between host infection patterns were more closely examined using
199 regression analyses of historical infection rates (weeks 20-40) to determine the strength
200 of these multivariate interactions. A linear model of chicken seroprevalence was limited
201 in its ability to account for variation in human incidence, yet was greatly improved when
202 examined as a non-linear relationship to account for the late-season decline in human
203 incidence while seroprevalence continued to increase (Fig. 1D). Similarly, a non-linear
204 analysis improved the ability of sentinel chicken seroprevalence to estimate variation in
205 *Culex* minimum infection rates (MIR) (Fig. 1E). Lastly, *Culex* MIR accounted for 71% of
206 the seasonal variation in human incidence when examined as a linear relationship (Fig.

207 1F). Together, these data suggest that both linear and non-linear relationships of host
208 infection rate data account for the dynamic nature of WNV activity. Traditionally used as
209 tools for arbovirus surveillance (48, 49), these data provide strong support for the
210 continued use of sentinel chickens and the use of mosquito infection rates to actively
211 monitor WNV activity.

212 **Regional differences in WNV transmission**

213 WNV infections were then examined across the state to identify spatial patterns of
214 endemicity and regions of higher risks for human infection. To approach this question,
215 mean annual infections rates were mapped for each host species (human, chicken,
216 mosquito) at the county level for all locations in which data have been collected from
217 2002 to 2016 (Fig. 2). Annual infection rates were also broken down by region, with
218 additional comparisons of infection intensity with respect to time (Fig. 2). From these
219 data, the human incidence of WNV was determined to be the highest in western Iowa,
220 where infection rates were four times that of central Iowa and seven times that of
221 eastern Iowa (Fig. 2A). Central Iowa also displayed significantly higher human incidence
222 than in eastern Iowa (Fig. 2A). Temporal analyses of human WNV incidence further
223 support this observation, demonstrating that incidence rates in western Iowa were
224 higher and amplified faster than in other geographic regions of the state (Fig. 2A).
225 These data confirm previous reports of WNV incidence in Iowa from 2002 to 2006 (15),
226 which similarly define a distinct gradient where the highest incidence rates are found in
227 western Iowa and the lowest incidence in eastern Iowa. This regional dichotomy
228 suggests that important entomological and ecological differences across the state must
229 influence higher rates of enzootic spillover of WNV in western Iowa.

230 Seroconversion rates in sentinel chickens were also highest in sentinel chickens in
231 western Iowa, displaying twice the seroconversion rates than those in central and
232 eastern Iowa (Fig.2B). However, unlike human incidence, these hotspots were less
233 defined with several non-western counties exhibiting high infection rates (Fig. 2B).
234 Seroconversion in western Iowa occurred earlier in the summer, resulting in infection
235 rates that were three times greater than those found in other regions by the end of the

236 year (Fig. 2B). No differences in seroconversion were detected between central and
237 eastern Iowa (Fig. 2B).

238 With similar spatial patterns of WNV in humans and sentinel chickens, we expected that
239 the MIR of *Culex* species producing WNV⁺ mosquito pools (Table S1) would likewise
240 display a similar pattern across the state. Although the majority of counties with the
241 highest infection rates were found in western Iowa, we found that the overall geographic
242 trends in mosquito infection rates were less pronounced (Fig. 2C). Mosquitoes from
243 western Iowa displayed a greater infection rate than those from central or eastern
244 regions, yet these differences were not significant (Fig. 2C). Temporal analyses
245 displayed equivalent amplification of mosquito infections in western and central Iowa,
246 with much lower mosquito infection rates in the east throughout the year (Fig. 2C).
247 While consistent with the geographic gradients for human incidence and chicken
248 seroprevalence, these mosquito infection trends argue that a more detailed analysis of
249 mosquito species composition on infection rates may be required to delineate potential
250 differences between *Culex* species as vectors of WNV transmission.

251 ***Cx. tarsalis* display higher infection rates than *Cx. pipiens* group mosquitoes**

252 Since the introduction of WNV into Iowa in 2002, *Cx. pipiens* group mosquitoes and *Cx.*
253 *tarsalis* have comprised 98% of the WNV⁺ mosquito pools (Table S1). To better
254 understand the contributions of these vectors to WNV transmission, we examined
255 mosquito infection data obtained from mosquitoes collected with gravid or CO₂-baited
256 CDC traps. Gravid traps lure mosquitoes seeking a site for oviposition and have been
257 widely used for monitoring populations of *Cx. pipiens* group species (50). Likewise,
258 mosquito yields from gravid traps were severely biased towards *Cx. pipiens* group
259 (99.5% of 93,456 collected; Fig. 3A). However, despite the sparsity of *Cx. tarsalis* data
260 from gravid trap collections, *Cx. tarsalis* displayed higher infection rates than *Cx. pipiens*
261 group mosquitoes (Fig. 3B).

262 Generally targeting mosquitoes seeking a blood meal (50), CO₂-baited CDC traps
263 resulted in more comparable yields of *Cx. tarsalis* and *Cx. pipiens* group mosquitoes
264 (Fig. 3C). *Cx. pipiens* group were still the more abundant vector (65.5% of 65,341
265 collected; Fig. 3C), but these data do not account for potential differences in species

266 distributions throughout the state. Comparisons of mean annual MIR between mosquito
267 species revealed that the infection rate in *Cx. tarsalis* was significantly higher (~5x) than
268 that of *Cx. pipiens* group mosquitoes (Fig. 3D). Therefore, *Cx. tarsalis* populations in
269 Iowa displayed higher infection rates than the *Cx. pipiens* group population during the
270 entire 15 year study period. This is further supported by laboratory and field-based
271 studies arguing that *Cx. tarsalis* is one of the most competent vectors of WNV in North
272 America (24, 51, 52).

273 **Blood meal analysis reveals differences in host selection between *Culex* vectors**

274 To examine potential differences in host selection between *Cx. tarsalis* and *Cx. pipiens*
275 group, we used PCR-based methods to analyze mosquito blood meals (30, 46, 53). Of
276 the mosquito samples analyzed, only a small percentage resulted in blood meal
277 identifications using our PCR-based methods (25% of *Cx. tarsalis* [23 of 92] and 20% of
278 *Cx. pipiens* group [32 of 159] samples). This is likely due to the passive collection of
279 these blood-fed mosquito samples in which host DNA in the blood meals may have
280 been digested past the point of PCR identification (54). These limitations prevent us
281 from making any strong assertions about differences in host preference between these
282 mosquito vectors. However, our data show that *Cx. tarsalis* feeds predominantly on
283 birds (~52% of all samples) and humans accounted for a large component (~30%) of
284 the identified blood meals (Fig. 4). This is in contrast to *Cx. pipiens* group mosquitoes
285 that collectively displayed stronger selection for non-human mammals and birds, with
286 humans representing the smallest proportion (~13%) of analyzed samples (Fig. 4).
287 When further classified by species using a multiplex PCR assay (44), we saw notable
288 differences within the *Cx. pipiens* group. *Cx. restuans* (which comprised the majority of
289 samples in our analysis) and *Cx. salinarius* fed mostly on non-human mammals (Fig.
290 S1), while *Cx. pipiens* samples fed primarily on birds (Fig. S1). Together, these data
291 suggest that *Cx. tarsalis* may exhibit different preferences for host selection than *Cx.*
292 *pipiens* group and feed more prevalently on humans (Fig. 4), providing support that *Cx.*
293 *tarsalis* may serve as a better vector of WNV to humans.

294 ***Cx. tarsalis* abundance is strongly correlated with human WNV infection**

295 Expanding on initial mosquito abundance data (15), we further examined *Culex*
296 mosquito populations across the state to determine the influence of vector abundance
297 on WNV transmission. Data collected from either CDC or New Jersey Light traps
298 (NJLTs) were used to calculate historical (2002 to 2016) *Culex* mosquito abundance
299 ratios used for comparisons to human WNV incidence on a county-specific level (Fig.
300 5). These data demonstrate that *Cx. tarsalis* comprises a larger proportion of the total
301 mosquito population in western Iowa than in other regions of the state (Figs. 5A and 5B;
302 15, 39, 55), and that its relative abundance significantly correlates with human WNV
303 incidence using either CDC (Fig. 5A) or NJLT trap datasets (Fig. 5B). In contrast, *Cx.*
304 *pipiens* group was widespread throughout the state, without clear associations to
305 human incidence (Figs. 5C and 5D). Together, these data demonstrate that human
306 incidence of WNV was highest in regions where *Cx. tarsalis* was most abundant. When
307 combined with its higher infection rates (Fig. 3) and greater selection for human hosts
308 (Fig. 4), these properties provide strong support towards incriminating *Cx. tarsalis* as
309 the principle vector of the majority of human WNV infections in Iowa.

310 Discussion

311 In this study, we provide a comprehensive view of 15 years of WNV transmission in
312 Iowa (2002 to 2016) following its introduction into the state. With the inclusion of
313 infection data from both vertebrate and invertebrate hosts, we provide a unique
314 perspective into the temporal and spatial trends that define WNV transmission in Iowa.
315 We describe a consistent period of WNV transmission in Iowa from late spring to early
316 fall, similar to patterns described in other locations of the Midwest (32, 35) and across
317 the country (9, 27, 56). All host species (human, bird, mosquito) show progressive virus
318 amplification throughout the summer months, with infection rates peaking in late
319 summer for humans and mosquitoes, followed by a dramatic decline with the onset of
320 fall. During this time, WNV remains prevalent in bird populations, which reach peak
321 levels of infection late into the summer and early fall. These transmission dynamics are
322 derived from 15 year averages, yet also display unique inter-annual differences in WNV
323 intensity. The mechanisms driving these infection patterns likely involve interrelated
324 complexities of physiology and behavior that influence interactions between hosts.

325 Shifts in vector feeding behavior have been previously implicated in driving WNV
326 transmission from an enzootic cycle to human populations (27), yet other studies
327 suggest that little change in feeding preferences occur with the emergence of human
328 cases (31). Therefore, it is unclear how these dynamics may influence human case
329 incidence in Iowa and the greater Upper Midwest region. Additional questions remain
330 regarding the factors that contribute to the dramatic decline of mosquito infection rates
331 and subsequent cessation of human disease cases in early October, despite the
332 presence of WNV in bird populations. This phenomenon is likely a combination of
333 declining temperatures and mosquitoes entering diapause, which would decrease
334 mosquito abundance and biting activity, thereby reducing interactions between vectors
335 and vertebrate hosts.

336 Similar to initial studies (15), we identify a distinct geographical gradient in the
337 prevalence of WNV across Iowa from east to west. The epidemiological importance of
338 the West is readily evident, where county-level incidence rates in western Iowa were as
339 much as 60-times higher than those in the East. Additional examination of sentinel
340 chicken seroprevalence and mosquito infection rates reinforce western Iowa as a region
341 of epidemiological importance for WNV. Due to the close proximity of Nebraska and
342 South Dakota, which harbor the highest human WNV incidence rates in the country (1),
343 our data imply that the western regions of Iowa may be influenced by similar landscape
344 factors driving WNV transmission as previously suggested (15).

345 Supported by more than 1.9 million individual mosquito records, we have accumulated a
346 significant body of evidence underscoring the importance of *Cx. tarsalis* in transmitting
347 WNV to humans and in shaping the hyper-endemic region of western Iowa. This is in
348 agreement with previous studies that have proposed important roles for *Cx. tarsalis* in
349 WNV transmission in the Upper Midwest and Northern Great Plains (14, 15, 34, 35).
350 With similar landscape ecology and strong agricultural emphasis across this region, the
351 widespread use of irrigation practices likely provide ideal habitats that enable *Cx.*
352 *tarsalis* to persist through periods of drought (15, 18). This is in contrast to *Cx. pipiens*
353 group species, which are commonly associated with peridomestic habitats throughout
354 the United States and have been implicated as the primary vector of WNV in urbanized

355 settings (9, 11, 12, 32). While we cannot exclude the involvement and contributions of
356 *Cx. pipiens* group in WNV amplification and human infections, the role of *Cx. pipiens*
357 group mosquitoes is likely that as a secondary vector for WNV transmission in regions
358 where *Cx. tarsalis* is abundant (summarized in Fig. 6). With Iowa serving as the
359 easternmost boundary of *Cx. tarsalis* distribution in North America (57), our data
360 suggest that Iowa serves as an important transition zone, in which *Culex* vector
361 importance shifts along a gradient from *Cx. pipiens* group mosquitoes in the eastern US
362 and Chicago (9, 11, 12, 32) to regions of high *Cx. tarsalis* abundance in Iowa and the
363 Upper Midwest. As a result, highlighting how differences in vector ecology shape
364 distinct epidemiological regions along an ecological gradient in Iowa provides a rare
365 opportunity to examine the species-specific dynamics of WNV transmission in multiple
366 mosquito vectors.

367 In addition, *Culex pipiens* group mosquitoes display distinct seasonality in their
368 abundance, evidenced by the earlier emergence of *Cx. restuans* before either *Cx.*
369 *pipiens* or *Cx. salinarius* (20, 58, 59). Previous studies have suggested that *Cx.*
370 *restuans* may be a comparable or better overall vector of WNV than *Cx. pipiens* (20,
371 60), such that early season populations of *Cx. restuans* may be integral in establishing
372 WNV amplification in reservoir populations before other vector species promote
373 enzootic spillover into human populations. Collectively referred to as *Cx. pipiens* group
374 in this study, additional efforts should be performed to evaluate the individual
375 contributions of *Cx. pipiens*, *Cx. restuans*, and *Cx. salinarius* to WNV transmission,
376 especially in the context of their spatiotemporal dynamics.

377 These differences in vector ecology and abundance are inherently linked to host
378 selection preferences between mosquito species that influence vector competence.
379 *Culex* species have generally been classified as ornithophilic, a preference that
380 promotes frequent interactions with bird reservoirs to acquire and amplify WNV. Several
381 studies have suggested that *Cx. pipiens* group principally feeds on birds (31, 46, 61,
382 62), while *Cx. tarsalis* is thought to be more opportunistic, feeding on both birds and
383 mammals (53, 62, 63). Our data support previous findings in Iowa (64) suggesting that
384 *Cx. tarsalis* fed on birds more frequently than mosquitoes of the *Cx. pipiens* group.

385 Moreover, our data suggest that *Cx. pipiens* group fed most commonly on mammals,
386 yet upon further speciation indicate that *Cx. restuans* and *Cx. salinarius* may have
387 stronger preference for non-human mammals while *Cx. pipiens* feed predominantly on
388 birds. *Culex* vectors have been shown to shift increasingly toward mammals through the
389 progression of summer (53, 63–65), a phenomenon that potentially drives epizootic
390 transmission (27), yet the limited scope of our study does not enable temporal analyses.
391 However, it is notable that *Cx. tarsalis* fed on humans more frequently than *Cx. pipiens*
392 group, a potential indicator of greater human host selection by *Cx. tarsalis* that would
393 support a larger role of this vector species in the transmission of WNV to humans. In
394 agreement with these studies, the data presented here support a model in which *Cx.*
395 *pipiens* group mosquitoes predominantly drive enzootic amplification of WNV in
396 reservoir populations, while *Cx. tarsalis* is more likely to transmit WNV to human
397 populations (Fig. 6).

398 In summary, our results demonstrate the power and utility of long-term mosquito
399 surveillance to provide a comprehensive understanding of WNV epidemiology and
400 transmission. Integrating human case data, sentinel chicken seroprevalence, and
401 mosquito infection rates, we provide an unprecedented amount of infection data that
402 enable us to define when and where WNV transmission is most active in the state of
403 Iowa. Through these studies, we present strong evidence implicating *Cx. tarsalis* as the
404 primary vector of WNV in Iowa, and identify a unique gradient in vector ecology that
405 shapes virus transmission in Iowa and the Upper Midwest region. With long-term WNV
406 surveillance at the local or state-wide level in many places throughout the country, often
407 with years of unpublished infection and mosquito data, we believe that our study can
408 serve as a template for similar long-term analyses to better understand WNV
409 transmission at the regional or state-wide level. Taken together, our results provide an
410 improved understanding of WNV transmission dynamics in an important epidemiological
411 region of North America.

412 **Author Contributions**

413 Conceived and designed the experiments: BMD RCS. Contributed data: BMD KBK EJJ
414 ENF WAR LCB RCS. Performed blood meal analysis: EJJ ENF. Analyzed the data:

415 BMD KBK EJG RCS. Funding and program oversight: WAR, LCB, RCS. Wrote the
416 initial draft of the manuscript: BMD KBK RCS. Edited the manuscript: BMD LCB RCS.
417 All authors read and approved the final manuscript.

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612 **Figure legends**

613 **Figure 1. Seasonality of WNV transmission in human, avian, and mosquito hosts.**
614 WNV infection data from Iowa (2002-2016) were analyzed and displayed as weekly
615 values for individual years or as the annual mean of human incidence (A), sentinel
616 chicken seroprevalence (B), or *Culex* minimum infection rate (MIR) (C). Correlations
617 between different host (human, avian, mosquito) infection rates were examined by
618 plotting averaged weekly values (weeks 20-40; data points on graph) to determine the
619 strength of host interactions involved in WNV transmission (D-F). A linear regression
620 (black dashed line; $R^2=0.45$, $P < 0.001$) or a best-fit 4th order polynomial (green line;
621 $R^2=0.94$, $P < 0.0001$) were used to explain sentinel chicken seroprevalence correlations
622 to human incidence (D). Similar linear regression (black dashed line; $R^2=0.63$, $P < 0.001$)
623 and best-fit 4th order polynomial (purple line; $R^2=0.86$, $P < 0.0001$) were performed to
624 examine the effects of *Culex* MIR in estimating sentinel chicken seroprevalence (E).
625 The non-linear relationship best explains the declining human incidence or mosquito
626 MIR, while sentinel chicken seroprevalence remained high at the end of the season (E).
627 Linear regression (orange line; $R^2=0.71$, $P < 0.0001$) analysis of *Culex* MIR and human
628 WNV incidence (F).

629 **Figure 2. Spatial distributions of WNV in human, avian, and mosquito hosts.**
630 WNV infections from 2002 -2016 are displayed at the county level and spatially
631 presented as the annual mean of human incidence (A), sentinel chicken seroprevalence
632 (B), or *Culex* MIR (C) across the state. Additional analyses display the annual mean or
633 weekly infection rates by regional associations in Iowa – western, central, and eastern
634 (A-C). Human incidence (A) is displayed as the number of human WNV cases per
635 100,000 people. Sentinel chicken seroprevalence (B) is defined as the ratio of WNV⁺
636 seropositive birds per total birds sampled. *Culex* MIR (C) is calculated as the minimum
637 infection rate for all *Culex* mosquito vectors (*Cx. pipiens* group, *Cx. tarsalis*, and *Cx.*
638 *erraticus*). The black dots display counties in which chicken seroprevalence and
639 mosquito infection rate data were collected. A Wilcoxon matched-pairs signed rank test
640 was used to test for differences between groups with significance denoted by asterisks
641 (* $P < 0.05$, ** $P < 0.01$, *** $P < 0.001$). ns, not significant.

642 **Figure 3. WNV surveillance reveals mosquito trapping bias and greater infection**
643 **rates in *Cx. tarsalis*.** (A) Gravid trap yields of *Cx. tarsalis* (red) and *Cx. pipiens* group
644 (blue) show significant trap bias toward collecting container-breeding *Cx. pipiens* group
645 mosquitoes. Mean annual minimum infection rates (MIR) of *Cx. tarsalis* and *Cx. pipiens*
646 group collected from gravid traps (B). Only two years of substantial *Cx. tarsalis* yields
647 were collected as a result of this trapping bias, resulting in a large margin of error (B). In
648 contrast, CDC trap yields of *Cx. tarsalis* (red) and *Cx. pipiens* group (blue) showed a
649 more even sampling of the two vectors (C), in which the mean annual MIR of *Cx.*
650 *tarsalis* were significantly higher in CDC trap collections (D). The Wilcoxon matched-
651 pairs signed test was used to test for differences between groups. Significant
652 differences are indicated by asterisks (** $P < 0.01$). ns, not significant.

653 **Figure 4. Identification of mosquito blood meals.**

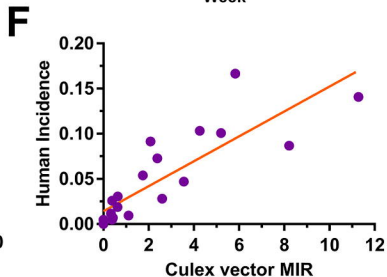
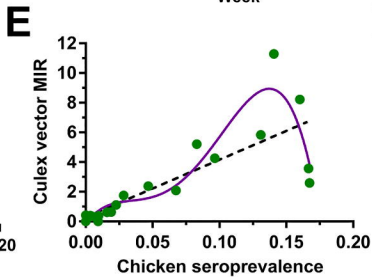
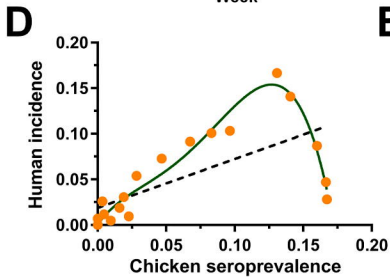
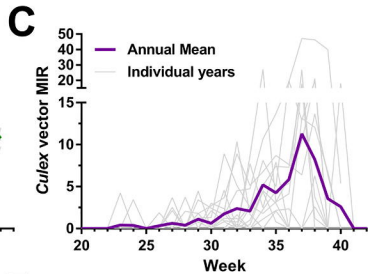
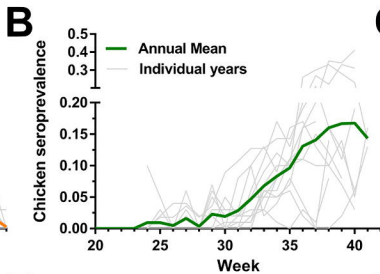
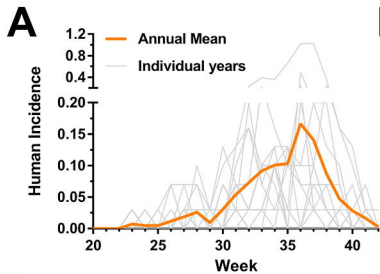
654 PCR-based methods were used to identify the blood meal source of collected *Cx.*
655 *tarsalis* and *Cx. pipiens* group samples. The percentage of blood meals taken
656 respectively from birds, humans, and non-human mammals are displayed for each
657 mosquito species. The total number (n) of identified blood meals is displayed below.

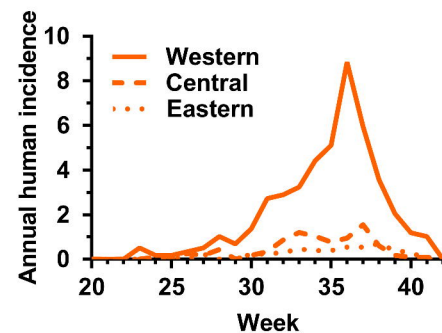
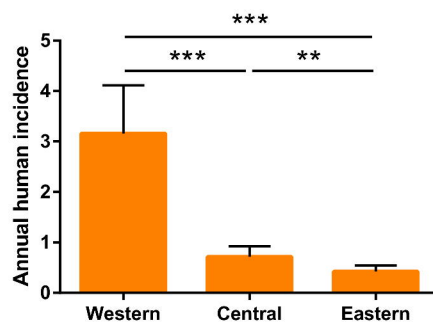
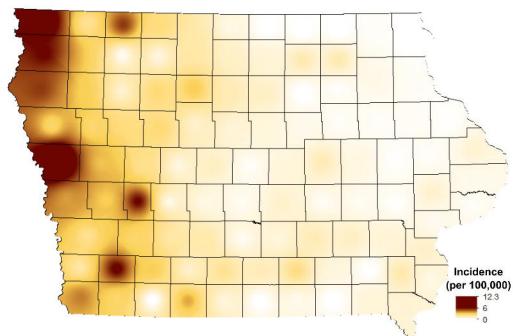
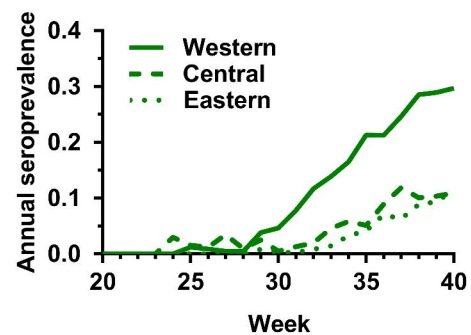
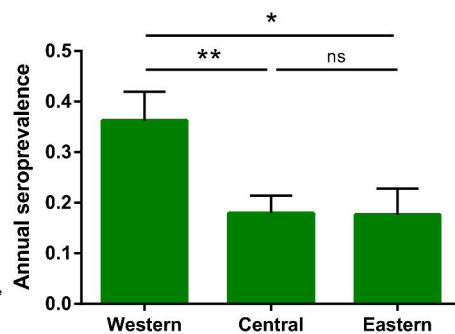
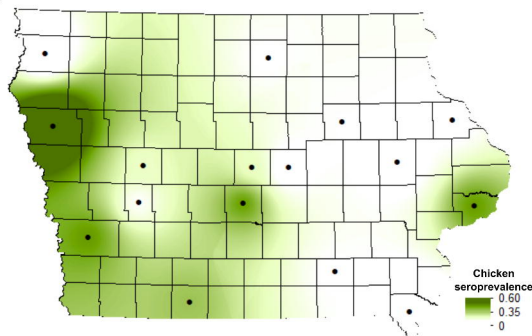
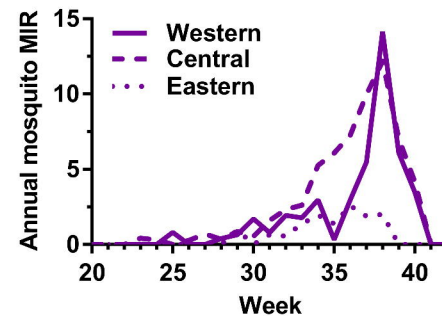
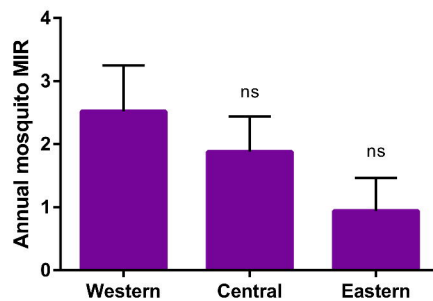
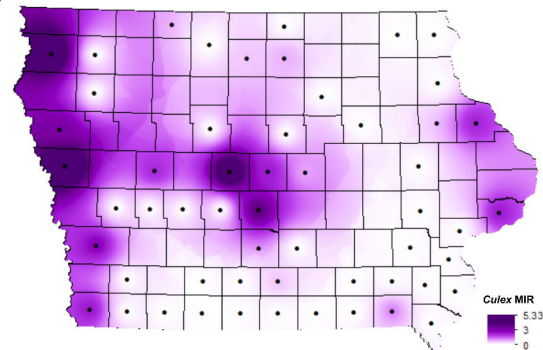
658 **Figure 5. *Cx. tarsalis* abundance significantly correlates with human WNV**
659 **incidence.** Mosquito abundance data collected from CDC traps (A and C) or New
660 Jersey light traps (NJLTs) (B and D) were examined spatially across Iowa. County-
661 specific mosquito datasets were calculated from historical data (2002-2016). Mosquito
662 abundance was measured as a ratio of either *Cx. tarsalis* (A and B) or *Cx. pipiens* group
663 (CPG) (C and D) mosquitoes to the total yield of mosquitoes (all species) collected.
664 Each figure shows a county-specific visual display of mosquito abundance in the state
665 (left) and the relationship of abundance values to human WNV incidence (right). Linear
666 regressions were performed to test the strength of the relationship between mosquito
667 abundance and human WNV incidence, with associated R^2 and P - values displayed for
668 each correlation.

669 **Figure 6. Model of WNV transmission dynamics in Iowa.**

670 *Culex* mosquito populations are believed to maintain and amplify WNV in bird
671 reservoirs, with spillover events into human populations primarily occurring through *Cx.*


672 *tarsalis*. *Cx. pipiens* group mosquitoes are believed to have a secondary role in human
673 cases.



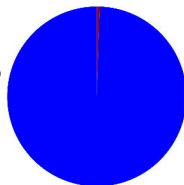
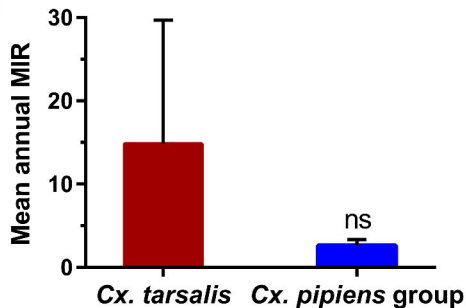
A**B****C**

A


Gravid trap

 *Cx. tarsalis* *Cx. pipiens* group

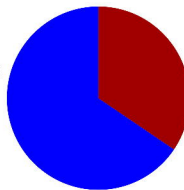
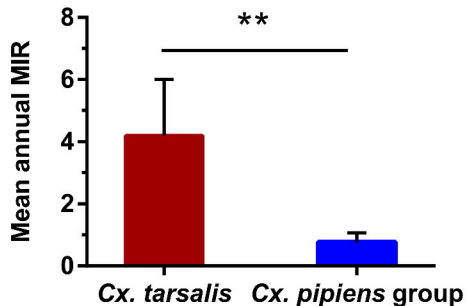
Total = 93,456

**B****C**

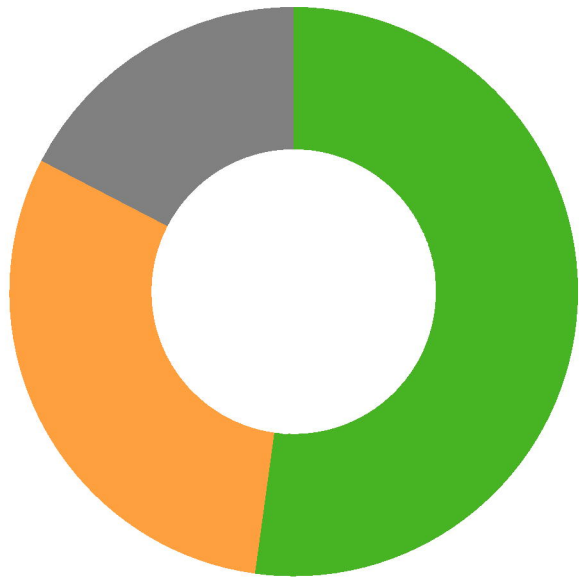
CDC trap

 *Cx. tarsalis* *Cx. pipiens* group

Total = 65,341

**D**

Cx. tarsalis



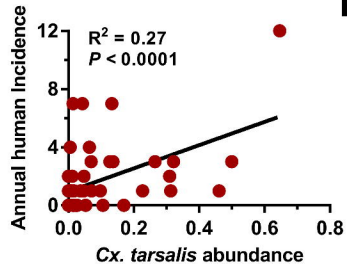
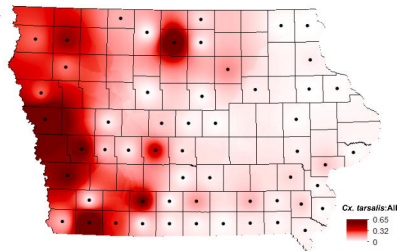
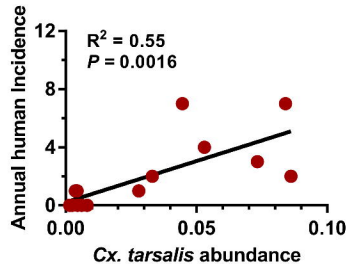
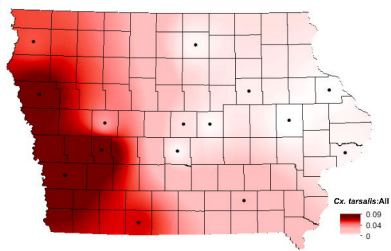
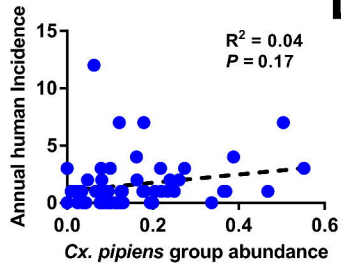
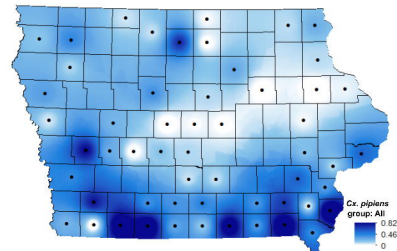
Total=23

Cx. pipiens group



Total=32

- Bird
- Human
- Mammal

A**B****C****D**