

1 **Title:** Wetland birds in the northern prairie pothole region may show sensitivity to agriculture.

2 **Authors:** Jody Daniel, Heather Polan and Rebecca C Rooney*

3 ***Corresponding Author:** B2-251 200 University Avenue West, Waterloo, ON, N2L3G1;

4 j9daniel@uwaterloo.ca

5 **Author Affiliations and Email Addresses:**

6 Jody Daniel: University of Waterloo, Department of Biology. j9daniel@uwaterloo.ca .ORCID:

7 0000-0003-3153-8164

8 Heather Polan: University of Waterloo, Department of Biology. hmpolan@gmail.com

9 Rebecca C Rooney: University of Waterloo, Department of Biology.

10 rebecca.rooney@uwaterloo.ca. ORCID: 0000-0002-3956-7210

11 **Abstract:**

12 Wetland losses in the Northern Prairie Pothole Region (NPPR) are largely attributed to
13 agriculture. Since land-use is known to influence bird habitat selection, bird community
14 composition is likely sensitive to the extent of neighboring agricultural activity. We determined
15 which local and landscape habitat variables are most predictive of wetland bird assemblage
16 occurrence in southern Alberta. We: 1) identified distinct bird assemblages with a cluster
17 analysis, 2) identified which species were indicative of these assemblages using an indicator
18 species analysis and 3) predicted which bird assemblage would occur in a wetland with a
19 classification and regression tree. Avian assemblages were more loosely defined and had few
20 indicator species. Importantly, assemblages were specific to the natural region in which the
21 wetland occurred. Also, landscapes with higher agricultural activity generally supported
22 waterfowl and shorebirds, likely because agricultural activities excluded wetland-dependent
23 birds that nest in upland habitat. Though waterfowl and shorebirds show poor sensitivity to

24 surrounding landscape composition, edge-nesting wetland avifauna may make good indicators of
25 ecological integrity.

26 **Introduction**

27 The majority of wetland losses in the Northern Prairie and Parkland Region (NPPR) are
28 attributed to agricultural and urban development (Kennedy and Mayer 2002; Mitsch and
29 Gosselink 2015), with agriculture leading to losses of about 90% of historic wetlands by 1951 in
30 the Canadian NPPR (Bethke and Nudds 1995). Wetlands lost to agriculture are usually filled or
31 drained to protect neighboring croplands from flooding and to increase cropland area (Schindler
32 and Donahue 2006; Verhoeven and Setter 2010). The remaining wetlands undergo physical and
33 chemical alterations (Rashford et al. 2011), which include 1) increased sedimentation due to
34 tillage (Zedler and Kercher 2005) and livestock grazing (Bloom et al. 2013); 2) higher nutrient
35 loading from fertilizer use (Schindler and Donahue 2006); 3) slower recovery rates when
36 exposed to disturbance (Bartzen et al. 2010); and 4) lengthened hydroperiods as soil infiltration
37 rates are lowered (van Der Kamp et al. 1999) and runoff is consolidated (McCauley et al. 2015).
38 Thus, wetlands that have escaped drainage or infilling may still be degraded by agricultural
39 activity in the surrounding landscape.

40 Even with conservation policies to protect wetlands, we continue to witness wetland loss
41 and degradation (Clare and Creed 2014; Davidson 2014). In the United States, for example, there
42 are no federal policies that manage farming practices around wetlands (Johnston 2014; Mitsch
43 and Gosselink 2015), though farmers require permits for activities that occur within a wetland
44 under Section 404 of the Clean Water Act (U.S. EPA 2017). Similarly, the Albertan wetland
45 policy (Government of Alberta 2013) offers legal protection to wetlands and introduces
46 innovation in shifting the focus of management from an area-basis to a function-basis. The

47 Albertan wetland policy fails to, however, provide protective buffers around wetlands
48 (Government of Alberta 2013). Consequently, despite legal protections and regulation of
49 activities that occur within wetland boundaries, wetland integrity and function may be
50 compromised by adjacent human activities through connections linking wetlands to their
51 catchments and beyond (Jones et al 2018; Kraft et al 2019).

52 Compromised wetland integrity may endanger bird populations because they are sensitive
53 to both changes in wetland condition and landscape structure. For example, Mensing et al.
54 (1998) found that, out of six taxa surveyed, birds were the best indicator of landscape condition
55 surrounding small-stream riparian wetlands. Bird diversity and richness were highly correlated
56 with the extent of cultivated land, wetland and forest cover within 500 and 1000 metre (m) radii
57 (Mensing et al. 1998). These findings are echoed in research in Alberta’s Parkland region, which
58 concluded that bird community integrity in shallow open-water wetlands was sensitive to road
59 density, forest cover, and the amount of other wetland habitat within 500 m (Rooney et al. 2012).

60 Most research on the drivers of bird composition in wetlands have focused on
61 permanently-ponded wetlands. Yet, temporarily- to semi-permanently-ponded wetlands also
62 comprise high-value bird habitat, especially for breeding and brood-rearing birds (Burger 1985;
63 Hands et al. 1991). Small, isolated wetlands sustain metapopulations (Semlitsch and Bodie
64 1998), and they are invaluable habitat for terrestrial, facultative, and obligate birds because 1)
65 there are lower occurrences of mammalian predators (Burger 1985); 2) there are interspersions of
66 mudflats that allow birds to dabble (Osborn et al. 2017), which allows them to feed while
67 remaining alert for predators (Pöysä 1986); 3) macroinvertebrate prey are abundant and diverse
68 (Zimmer et al. 2000; Gleason and Rooney 2017); and 4) the absence of fish improves food
69 availability for birds (Zimmer et al. 2001). For example, Shealer and Alexander (2013) reported

70 that insectivorous Black Terns (*Chlidonias niger*), which nest in more-permanently flooded
71 wetlands, commonly forage in temporarily-flooded wetlands up to 4 kilometres (km) away.
72 Since wetlands that dry up during the breeding season provide additional foraging opportunities
73 for birds and refuge from predators, they are valuable bird habitat.

74 Since land-use is known to influence bird habitat selection (Ballard et al. 2014), we
75 anticipate that bird community composition and guild structure in prairie pothole wetlands will
76 be sensitive to the extent of agricultural activity in adjacent lands. We seek to determine which
77 local- and landscape-level habitat variables are most predictive of bird assemblage occurrence in
78 prairie potholes of Alberta. If birds are sensitive to agricultural activity in the surrounding
79 landscape, it raises concerns that existing wetland policy that fails to provide buffer protections
80 surrounding wetlands, may fail to protect bird communities and the important ecological services
81 they provide. Furthermore, we evaluate the dependency of these predictions on wetland-
82 dependent birds, including shorebirds, wetland-dependent songbirds and waterfowl. We asked 1)
83 if there are distinct assemblages of birds occupying these wetlands, 2) if so, what habitat traits at
84 the local- and landscape-level are predictive of assemblage occurrence; and 3) whether bird
85 assemblages could be used to indicate the level of agricultural disturbance affecting a prairie
86 pothole wetland.

87 **Methods**

88 *Study Area*

89 Our study region encompasses the Parkland and Grassland natural regions of Alberta
90 (Fig. 1). In this semi-arid climate, evapotranspiration rates exceed annual precipitation (Downing
91 and Pettapiece 2006; Millett et al. 2009), but depressions created by glaciation nonetheless give
92 rise to a high density of small wetlands known as prairie potholes. The Parkland is cooler and

93 moister, supporting a mosaic of deciduous forest and grassland. The Grassland is warmer and too
94 dry for most trees (Downing and Pettapiece 2006).

95 ***Study Design***

96 We selected 72 natural wetlands than ranged in permanence class from temporarily-
97 ponded to permanently-ponded (*sensu* Stewart and Kantrud 1971) and were evenly distributed
98 among six, randomly selected sub-watersheds (three in each Natural Region – Grasslands and
99 Parklands) of glaciolacustrine or glaciofluvial-derived surficial geology (Fig. 1). Our selected
100 wetlands reflected the frequency distribution of wetland sizes within each sub-watershed, based
101 on the Alberta Merged Wetland Inventory (Government of Alberta 2014a), and so were
102 generally small (mean size $0.66 \pm \text{SE } 0.07$ ha). To guard against spatial dependencies, sites were
103 spaced a minimum of 3.5 km apart. Independent of their hydroperiod, wetlands were selected to
104 span a gradient in the extent of agricultural activity in the surrounding landscape (i.e., the
105 percentage cropping, haying, and pastureland covers within 500 m buffers around each wetland's
106 perimeter). Land cover data were derived from the Agriculture and Agri-Food Canada Annual
107 Crop Inventory Data (AAFC 2015) and supplemented with information from the provincial
108 Grassland Vegetation Inventory (Government of Alberta 2014b). We used a 500 m buffer
109 because this scale has been reported as the most influential of bird community integrity in
110 permanently-ponded wetlands in Alberta's Parkland (Rooney et al. 2012). Due to the level of
111 agricultural activity in the Grassland natural region (Downing and Pettapiece 2006), where over
112 70% of land is privately owned (AEP 2011), truly pristine reference sites are scarce.
113 Consequently, we classified wetlands with less than 25% agricultural land cover as being in the
114 least disturbed condition (*sensu* Stoddard et al. 2006) and used these as a reference condition

115 against which medium (25-75% agricultural cover) and high disturbance (greater than 75%
116 agricultural cover) wetlands could be compared.

117 *Bird Surveys*

118 Bird surveys were conducted by pairs of observers in 2014 and 2015, following the
119 method described in Wilson and Bayley (2012). In brief, surveys comprised a 10-min visual
120 survey followed by an 8-minute acoustic fixed-radius point count survey, with a radius of 100 m.
121 Because most wetlands were less than 1 ha, a single 100-m radius point count covered the entire
122 wetland. Larger sites were surveyed from two-point count locations, providing they could be
123 positioned at a minimum of 200 m apart, in which case counts were summed to reflect the
124 wetland as the sample unit.

125 Surveys were conducted twice at each wetland during the breeding season (May 19th –
126 June 24th) to account for any temporal partitioning of breeding activity among species within the
127 general breeding season. Consequently, we summed counts across surveys. Generally, birds in
128 our study region sing and call between sunrise and 11:00 am (Farr et al. 2012). Thus, all surveys
129 were restricted to this time period.

130 All birds visually observed foraging or nesting or heard singing or calling at the site were
131 enumerated and identified to species. Bird identifications followed the American Ornithologists
132 Union Standard Information used to determine guild membership of bird species, such as feeding
133 traits, preferred habitat, and migration patterns, were retrieved from Birds of North America
134 Online (CLO 2014). We distinguish between this complete bird assemblage and the subset of
135 birds observed using the marsh that are classified as wetland-dependent species (Online
136 Resource 1); only these wetland-dependent species were included in our subsequent analyses.

137 ***Local-level Habitat Characterization***

138 We surveyed the vegetation at each wetland during peak aboveground biomass between
139 late July and August. First, we used a sub-meter accuracy GPS (Juno Trimble T41; SXBlue II
140 GPS/GNSS Receiver) to delineate the wetland boundary such that the perimeter of the wetland
141 lay where vegetation transitioned to <50% cover by wetland-obligate plant species. Next, we
142 sub-divided the wetland into zones based on vegetation form (woody vegetation, drawdown,
143 ground cover, narrow-leaved emergent, broad-leaved emergent, robust emergent, open-water
144 area) and the associated dominant or co-dominant macrophyte species. These vegetation zones
145 were delineated in the same manner as the wetland and their area calculated in the field to inform
146 quadrat-based sampling intensity. Each vegetation zone was then characterized by a minimum of
147 five 1 m² quadrats. If a zone was larger than 5000 m², we added one quadrat per 1000 m².
148 Finally, we estimated the mean percent cover among plots and then relativized our estimates to
149 100%, for a site-level estimate of vegetation cover.

150 In addition to vegetation surveys, we monitored abiotic variables known to influence
151 bird habitat selection. From May and September 2014, we measured water depth using staff
152 gauges, providing ponded water remained in the wetland. This was used to calculate the
153 wetland's maximum water, minimum water depth, and seasonal amplitude (maximum depth
154 minus minimum depth).

155 ***Statistical Analysis***

156 Our analysis objectives were 1) to test whether the birds grouped into distinct
157 assemblages; and 2) to determine which local- and landscape-level variables were most
158 predictive of bird assemblage occurrence by developing a model to classify wetlands in terms of

159 their expected bird assemblage based on local- and landscape-variables, with particular emphasis
160 on the level of agricultural activity surrounding each wetland.

161 We used a square-root transformation and relativized our wetland-dependent bird count
162 data by the maximum value in each column to improve multivariate normality and reduce the
163 influence of numerically-dominant species. To reduce data sparsity, we removed rare avian
164 species (<2 occurrences out of 72 wetlands). Following the recommendations of McCune and
165 Grace (2002), we used a Bray-Curtis dissimilarity measure to characterize distances in species
166 space of community composition among our wetlands.

167 *Cluster & Indicator Species Analysis*

168 We used a cluster analysis to identify distinct wetland-dependent bird assemblages
169 among our sites. We used a hierarchical agglomerative polythetic process for the cluster analysis
170 using the cluster package (Maechler et al. 2018) in R statistical software (R Core Team 2017).
171 For the cluster analysis, we specified a flexible beta linkage method (beta = -0.250) and used the
172 Bray-Curtis distance measure, based on the recommendations of McCune and Grace. (2002).
173 Also using the cluster package, we then pruned the dendrogram iteratively, varying the number
174 of groups among sites from two to 20.

175 We used an indicator species analysis (ISA) to determine the optimal number of wetland-
176 dependent bird assemblages among our sites as the number of groups generating the smallest
177 average p -value across indicator species. As described in Dufrêne and Legendre (1997), this
178 analysis estimates the indicator value of each species based on their relative abundance and
179 frequency in each group and assigns a measure of statistical significance using a Monte Carlo
180 method with 4999 permutations. For the ISA, we used the labdsv package in R (Roberts 2016),
181 and the site-group memberships (overall number of assemblages among our sites) derived from

182 the trimmed dendrogram. In ISAs, because groups with one sample unit must be excluded from
183 analysis (Peck 2010), we limited our analysis to site-group memberships with at least two sites
184 per group.

185 *Visualizing Community Composition*

186 To visualize how wetland-dependent bird communities are related to the local and
187 landscape variables, we ran nonmetric multidimensional scaling ordinations (NMDSs) on our
188 two bird matrices. We used the NMDSs to visualize how 1) the local and landscape variables in
189 the final classification and regression tree (CART) model (described below) were related to
190 community composition, and 2) functional traits were related to each wetland-dependent bird
191 assemblage identified in the ISA. We used the vegan package to implement the NMDSs
192 (Oksanen et al. 2017) in R statistical software (R Core Team 2017).

193 After implementing each NMDS, we used vector overlays to visualize how species
194 counts ($r^2 > 0.2$ with at least one axis) and counts of species possessing various functional traits
195 aligned ($r^2 > 0.1$ with at least one axis) with major trends in bird community composition. We
196 symbolized sites by assemblages identified in the combined cluster analysis and ISA.

197 *Classification and Regression Tree*

198 Finally, we developed a classification and regression tree to predict which wetland-
199 dependent bird assemblage would occur at a marsh, using a combination of local- and landscape-
200 level data. In our case, the classification and regression tree partitions the wetlands based on
201 local- and landscape-level characteristics to create nodes of wetlands such that the deviance
202 between node membership and bird assemblage cluster is minimized. We used local-level (size,
203 percentage cover of woody, robust emergent and broad leaf plants, maximum water depth) and
204 landscape-level variables (percentage cover of grassland, forest and shrubs, water and wetlands,

205 cropland and human-related land use within a 500 m radius) that would be critical in influencing
206 the functional traits of wetland-dependent birds present in a wetland, as predictors in the
207 classification tree.

208 We used the “tree” package (Ripley 2016) in R statistical software (R Core Team 2017),
209 to implement the classification and regression tree. The classification tree implements binary
210 recursive partitioning, using the deviance index described in Breiman et al. (1984) to estimate
211 impurity for splitting, and stops splitting when the terminal node passes a size threshold for the
212 number of wetlands included (Ripley 2016). Next, we used k-fold cross-validation to prune the
213 tree, where $k = 10$, which was based on cost-complexity as measured by deviance. We also used
214 the “tree” package to determine the number of misclassifications for the overall tree, as well as
215 the number of misclassifications at each node. Because our small sample size could contribute to
216 unstable k-fold cross-validation errors with increasing tree size, we repeated the test 100 times
217 and found the mean and standard error across iterations.

218 We used goodness of fit tests to measure if our classification and regression tree
219 predictions differed from the groups generated by the combined cluster analysis and ISA. Using
220 the DescTools package (Signorell 2017) in R, and a Williams correction for our small sample
221 size, we performed a G-Test. Next, we used the caret package (Kuhn 2017) in R to examine
222 whether there was strong agreement between our classification and regression tree predictions
223 and ISA assemblages, using kappa statistics.

224 **Results**

225 *Cluster & Indicator Species Analysis*

226 We differentiated five distinct wetland-dependent bird assemblages (dendrogram in Fig. 2;
227 indicator values listed in Table 1), using agglomerative hierarchical clustering and ISA. We

228 assigned each assemblage a name reflecting the life history traits of the birds that were the
229 strongest indicators of the assemblage (indicator values listed in Table 1). Only a few species
230 were considered significant indicators of the five wetland-dependent bird assemblages. Note that
231 all but one wetland-dependent bird assemblage had at least one significant indicator species that
232 was both faithful and relatively exclusive to that assemblage of birds. The exception is the
233 Hummock Nesters (Table 1), which was the first assemblage to merge with another (Pond and
234 Reed Associates) at a Bray-Curtis dissimilarity value of about 0.8 during cluster analysis. The
235 strongest indicator species for the Hummock Nesters was Wilson's Snipe (*Gallinago delicata*),
236 with an indicator value of 25.35 ($p = 0.100$). A list of indicator values of all bird species included
237 in the cluster and indicator species analyses is presented in Table 1.

238 ***Visualizing Community Composition***

239 Based on an assessment of the marginal decline in stress with increasing dimensionality,
240 we concluded that a three-dimensional solution was optimal for both our NMDS ordinations. The
241 NMDS, final stress was 18.58, and the NMDS converged in fewer than 20 runs.

242 The abundance of wetland-dependent birds differed among assemblages, based on their
243 nesting or habitat preferences (Fig. 3). Unsurprisingly, the Shrub Associates assemblage
244 supported shrub species [e.g. Gray Catbird (*Dumetella carolinensis*)]. However, both shorebird
245 [e.g. Willet] and non-shorebird species [e.g. Mallards (*Anas platyrhynchos*)] were associated
246 with the Wetland Edge Nester assemblage (Fig. 3C; 3D). Wetlands classified as supporting the
247 Shorebird Assemblage contained abundant shorebird and ground nesting species (Fig. 3A; 3B).
248 The Hummock Nester-classified wetlands shared species with all assemblages except the shrub
249 associates (Fig. 3A; 3B), and they were not strongly associated with any bird nesting or habitat

250 preferences (Fig. 3C; 3D). Conversely, only marsh (e.g. Sora) or pond species (e.g. American
251 Coot) were associated with the Pond and Reed Nesters assemblage (Fig. 3A; 3C).

252 The NMDS axes were indicative of various local- and landscape-scale wetland
253 characteristics. Axis one in the NMDS reflected a disturbance gradient (Fig. 3C; 3D), wetlands
254 differed in which Natural Region they were located along axis two (Fig. 3C) and axis three
255 reflected a hydroperiod gradient (Fig. 3D).

256 *Classification and Regression Tree*

257 *All Birds*

258 Using a combination of local and landscape-level variables (comprehensive list of
259 variables in Online Resource 2), we predicted which of the bird assemblages would occur at a
260 given wetland. The classification tree had ten terminal nodes (Fig. 4), with low total residual
261 deviance (12.48) and a misclassification error rate of 27%. Based on 10-fold cross validation
262 error, we trimmed the tree from ten (cross-validation error = 60%) to eight (cross-validation error
263 = 59%) terminal nodes (Table in Online Resource 3). The pruned tree had a marginally higher
264 total residual deviance (12.94), but the same misclassification error rate (27%).

265 The pruned model predicted all five assemblages. The model predictions had strong
266 agreement with the assemblages from the ISA ($\kappa = 66\%$). More, any differences between the
267 classification tree predictions and the observed assemblages were not statistically significant (G-
268 Test: $G = 10.19$, $df = 63$, $p\text{-value} = 1.00$).

269 The Wetland Edge Nesters and Shorebirds assemblages were the most distinct (in local
270 and landscape characteristics), occurring in a single terminal node (Fig. 4). The other
271 assemblages each occurred in two terminal nodes but differed in the distances between nodes
272 (Fig. 4). The Shorebirds assemblage was the third-most distinct assemblage, though predicted at

273 different tree heights. The Pond and Reed Nesters had the furthest vertical distance between the
274 nodes. However, the Hummock Nester-classified wetlands were predicted in both regions,
275 suggesting they were the least distinct assemblage.

276 Misclassification error rates were moderate (0 – 46%) (Table 2). Error rates were highest
277 for adjacent assemblages (e.g. Wetland Edge Nesters vs. Hummock Nesters), supporting birds
278 with similar foraging and nesting preferences. Conversely, error rates were low for assemblages
279 that were restricted to a region (e.g. Shorebirds vs Pond and Reed Nesters) (Fig. 4; Table 2).

280 Local wetland characteristics were most predictive of assemblages (Fig. 4). Similar to our
281 analysis on all birds using a wetland, Natural Region was the strongest predictor of assemblages.
282 Apart from Hummock Nesters and Wetland Edge Nesters, the assemblages were predicted by
283 proxies for wetland hydroperiod (e.g., size, depth) and vegetation characteristics (e.g., robust
284 emergent vegetation cover, broadleaf vegetation cover).

285 **Discussion**

286 Although wetland policy across North America aims to protect wetlands, neither
287 American nor Canadian policy limits what activities can take place in the immediate landscape
288 surrounding wetlands. If wetland bird communities are sensitive to adjacent land cover/land use,
289 then existing policy may be incapable of conserving wetland functions without incorporating
290 some buffer protections. Conversely, if avifauna are sensitive only to in situ wetland conditions
291 such as hydrology and vegetation structure, then buffer protections should not be necessary to
292 conserve bird community function and integrity. Based on prior research (e.g., Rooney et al
293 2012; Anderson and Rooney 2019), we predicted that bird community composition and guild
294 structure in wetlands of the Grassland and Parkland Natural Regions of Alberta would be
295 sensitive to the proportion of agricultural land cover in the surrounding landscape. We found

296 some support for this prediction – one of our five wetland-dependent bird assemblages were
297 absent from wetlands in agriculturally-dominated landscapes. We attribute this to differences in
298 the nesting behaviors of this assemblage – they nested in upland habitat and consequently
299 selected for landscapes with more natural land cover, mainly grassland. Conversely, dabblers,
300 divers, and waders were indicative of assemblages in landscapes with more agricultural activity
301 and longer hydroperiods. Consequently, waterfowl and shorebirds are less sensitive to this land
302 conversion, and so come to dominate wetlands situated in agricultural landscapes. However, the
303 importance of surrounding land cover in predicting which wetland-dependent assemblage would
304 occur at a wetland was less evident than we anticipated.

305 The most significant predictor of bird assemblage occurrence was the Natural Region that
306 the wetland fell in – Parkland vs. Grassland. The Grassland and Parkland natural regions differ in
307 both their landscape- and local-level characteristics. At the landscape-level, the Parkland
308 supports copses of aspen forest and more shrubland than the Grassland (Downing and Pettapiece
309 2006). Further, while there is more cropland in the Parkland, pastureland and haying are more
310 common in the Grassland (Downing and Pettapiece 2006). Also, because of differences in
311 climate, we observe a higher abundance of wetlands with longer hydroperiods in the Parkland
312 (Government of Alberta 2014a). In the more arid Grassland, the magnitude of difference
313 between potential evapotranspiration and precipitation is larger, resulting in the dominance of
314 shorter-hydroperiod wetlands (e.g., temporary and seasonal). Thus, we likely find Natural
315 Region to be a strong predictor of avian assemblage occurrence because of the preference of
316 some bird species for shorter-hydroperiod wetlands in mixed-grass prairie typical of the
317 Grassland (e.g., Shorebirds) versus the preference of other bird species for longer-hydroperiod

318 wetlands in landscapes with more shrubland and forest typical of the Parkland (e.g., Pond/Reed
319 Associates).

320 The composition of the landscape surrounding a wetland is not strongly predictive of
321 which wetland-dependent assemblage we find occupying a wetland. Local-level factors such as
322 hydroperiod (wetland permanence class, depth and size) and vegetation characteristics
323 (percentage cover of broadleaf and robust emergent plants) were important predictors of
324 assemblage occurrence. An explanation for this influence of local-level factors is that both
325 wetland hydroperiod and vegetation dictate food availability (Lantz et al. 2011) and nesting
326 opportunities for wetland-dependent birds (Kantrud and Stewart 1984). Because many wetland-
327 dependent birds have feeding behaviors (e.g., diving, dabbling, and wading) tied to the
328 availability of open water habitat, these local-level factors were critical in determining whether
329 their habitat needs could be met. For instance, the Pond and Reed Associate and the Shorebird
330 assemblages were distinguished using both the all birds and wetland-dependent bird datasets.
331 These assemblages were characterized by birds that dive, dabble, and wade to feed. More, these
332 assemblages were predicted to occur in wetlands that were deeper, larger, and had longer
333 hydroperiods and nearly all their indicator species were ground, pond or reed nesters that nest in
334 the wetland proper. Thus, we conclude that these assemblages were most sensitive to in situ
335 factors about the wetland, rather than the character of its surrounding landscape.

336 Waterfowl and shorebirds may dominate wetlands in landscapes heavily influenced by
337 agriculture not necessarily because they profit from cropping and grazing activities, but because
338 species reliant on upland habitat for nesting are excluded. Both the Wetland Edge Nesters and
339 Hummock Nesters assemblages were predicted to occur in deeper (>0.53 m), larger (>10745 m²)
340 wetlands in the Grassland, but it was the Hummock Nester assemblage that occurred in wetlands

341 with higher cropland activity in the surrounding landscape (>42 %). Consequently, species
342 belonging to the Hummock Nester assemblage come to dominate these wetlands because their
343 nesting habitat is still available when upland habitat is lost to agriculture, while wetland birds
344 that typically nest in upland habitat are now unable to do so (e.g., species the Wetland Edge
345 Nester assemblage). Similarly, Anderson and Rooney (2019) reported that significant differences
346 in bird community composition between natural and restored wetlands in the Parkland region of
347 Alberta were only evident when all birds were considered. They also reported that any difference
348 in the composition of wetland-dependent birds were negligible because restored wetlands were
349 similar to natural wetlands in their size, hydroperiod, and vegetation zonation, but differed
350 significantly in terms of landscape context. Therefore, by using a more comprehensive bird
351 survey data, we can develop bird-based wetland monitoring and assessment tools that reflect the
352 community-wide impacts of land cover change on bird assemblage occurrence.

353 Our CART and NMDS can be useful tools in designing wetlands for wetland-dependent
354 birds. Though the species pool did differ between Natural Regions, landscape composition can
355 be important when designing wetlands for birds. For example, if a practitioner aimed to provide
356 habitat for a Shorebird assemblage in the Grassland Natural Region, the wetland should be deep
357 (> 0.53 m) or large (>10745 m²) (i.e., CART results) and have lower human activity (i.e., NMDS
358 results). However, if the said practitioner was targeting Wetland Edge Nesters, the wetland can
359 be smaller (<10745 m²) but should have moderate to low cropping activity in the landscape
360 (<42.9 %).

361 ***Conclusion***

362 We show that, generally, wetland-dependent assemblages show poor sensitivity to
363 agricultural activity. While waterfowl and shorebirds were sensitive to in situ properties of the

364 wetland, such as water depth and wetland size or vegetation zonation patterns, edge-nesting birds
365 were excluded from wetlands with higher cropping activity. Waterfowl and shorebirds seem to
366 dominate wetlands in landscapes with more agricultural activity because other avian species are
367 excluded, despite being at greater risk of predation in these landscapes (Emery et al. 2005).
368 SWhen designing wetlands for use by these wetland avifauna, our concurrent analyses using a
369 CART and NMDS are useful tools in determining the landscape context and wetland
370 characteristics suitable for assemblages that may be the target in restoration policy.

371 **Acknowledgements**

372 Funding for this research was provided by Alberta Innovates grant #2094A. We are
373 grateful to Dr. Derek Robinson who assisted with site selection and land cover analysis. We are
374 also grateful to Drs. Stephen Murphy and Roland Hall who provided feedback on an early draft
375 of this manuscript. We thank Daina Anderson, Brandon Baer, Matt Bolding, Graham Howell,
376 Adam Kraft, Jennifer Gleason and Nicole Meyers for collecting the field data. Finally, we thank
377 Dr. Erin Bayne for supplying the automated recording units, which were used to verify auditory
378 surveys.

379 **References**

- 380 AAFC (2015) Annual Crop Inventory - 2009-2014. Government of Canada:
381 <http://open.canada.ca/data/en/dataset/ba2645d5-4458-414d-b196-6303ac06c1c9>
- 382 AEP (2011) Green/White Area. Government of Alberta, Edmonton, Alberta:
383 [https://geodiscover.alberta.ca/geoportal/catalog/search/resource/details.page?uuid=%7BDF](https://geodiscover.alberta.ca/geoportal/catalog/search/resource/details.page?uuid=%7BDF54991B-D9DB-494F-B2E0-0A8D711D9CAB%7D)
384 [54991B-D9DB-494F-B2E0-0A8D711D9CAB%7D](https://geodiscover.alberta.ca/geoportal/catalog/search/resource/details.page?uuid=%7BDF54991B-D9DB-494F-B2E0-0A8D711D9CAB%7D)
- 385 Anderson DL, Rooney RC (2019) Differences exist in bird communities using restored and
386 natural wetlands in the Parkland region, Alberta, Canada. *Restoration Ecology* 27:1495–
387 1507. doi: 10.1111/rec.13015
- 388 Ballard T, Seager R, Smerdon JE, et al (2014) Hydroclimate variability and change in the prairie
389 pothole region, the “Duck factory” of North America. *Earth Interactions* 18:1–28. doi:
390 10.1175/EI-D-14-0004.1
- 391 Bartzan BA, Dufour KW, Clark RG, Dale Caswell F (2010) Trends in agricultural impact and
392 recovery of wetlands in prairie Canada. *Ecological Applications* 20:525–538. doi:
393 10.1890/08-1650.1
- 394 Bethke RW, Nudds TD (1995) Effects of climate change and land use on duck abundance in
395 Canadian prairie-parklands. *Ecological Applications* 5:588–600. doi: 10.2307/1941969
- 396 Bloom PM, Howerter DW, Emery RB, Armstrong LM (2013) Relationships between grazing
397 and waterfowl production in the Canadian prairies. *Journal of Wildlife Management*
398 77:534–544. doi: 10.1002/jwmg.497
- 399 Breiman L, Friedman J, Stone CJ, Olshen RA (1984) *Classification and regression trees*. Taylor

- 400 & Francis, Belmont, CA
- 401 Burger J (1985) Habitat selection in temperate marsh-nesting birds. *Habitat Sel. Birds*
- 402 Clare S, Creed IF (2014) Tracking wetland loss to improve evidence-based wetland policy
- 403 learning and decision making. *Wetlands Ecology and Management* 22:235–245. doi:
- 404 10.1007/s11273-013-9326-2
- 405 Cornell Laboratory of Ornithology (2014) The Birds of North America Online.
- 406 <http://bna.birds.cornell.edu/bna/>.
- 407 Davidson NC (2014) How much wetland has the world lost? Long-term and recent trends in
- 408 global wetland area. *Marine and Freshwater Research* 65:934–941. doi: 10.1071/MF14173
- 409 Downing DJ, Pettapiece WW (2006) Natural Regions and Subregions of Alberta. Government of
- 410 Alberta, Edmonton, Alberta
- 411 Dufrière M, Legendre P (1997) Species assemblages and indicator species: The need for a
- 412 flexible asymmetrical approach. *Ecological Monographs* 67:345–366. doi:
- 413 10.2307/2963459
- 414 Emery RB, Howerter DW, Armstrong M, et al (2005) Seasonal variation in waterfowl nesting
- 415 success and its relation to cover management in the Canadian prairies. *Journal of Wildlife*
- 416 *Management* 69:1181–1193. doi: 10.2193/0022-541X(2005)069[1181:SVIWNS]2.0.CO;2
- 417 Farr D, Shank C, Moses R, et al (2012) Terrestrial field data collection protocols (abridged
- 418 version), Version 20. Alberta Biodiversity Monitoring Institute, Edmonton, Alberta
- 419 Gleason JE, Rooney RC (2018) Pond permanence is a key determinant of aquatic
- 420 macroinvertebrate community structure in wetlands. *Freshwater Biology* 63:264–277. doi:

- 421 10.1111/fwb.13057
- 422 Government of Alberta (2013) Alberta Wetland Policy. Government of Alberta, Environment
423 and Sustainable Resource Development, Edmonton, Alberta
- 424 Government of Alberta (2014a) Alberta Merged Wetland Inventory.
- 425 Government of Alberta (2014b) Grassland Vegetation Inventory (GVI) □: a multi-species
426 conservation strategy for species at risk in the grassland natural region of Alberta.
427 Edmonton, Alberta
- 428 Hands HM, Ryan MR, Smith JW (1991) Migrant Shorebird Use of Marsh, Moist-soil, and
429 Flooded Agricultural Habitats. *Wildlife Society Bulletin* 19:457–464. doi: 10.2307/3782158
- 430 Johnston CA (2014) Agricultural expansion: land use shell game in the U.S. Northern Plains.
431 *Landscape Ecology* 29:81–95. doi: 10.1007/s10980-013-9947-0
- 432 Jones CN, McLaughlin DL, Henson K, et al (2018) From salamanders to greenhouse gases: does
433 upland management affect wetland functions? *Frontiers in Ecology and the Environment*
434 16:14–19. doi: 10.1002/fee.1744
- 435 Kantrud H a, Stewart RE (1984) Ecological distribution and crude density of breeding birds on
436 prairie wetlands. *The Journal of Wildlife Management* 48:426. doi: 10.2307/3801174
- 437 Kennedy G, Mayer T (2002) Natural and constructed wetlands in Canada: an overview. *Water*
438 *Quality Research Journal of Canada* 37:295–325. doi: 10.1111/j.1469-8137.2010.03543.x
- 439 Kraft AJ, Robinson DT, Evans IS, Rooney RC (2019) Concordance in wetland physicochemical
440 conditions, vegetation, and surrounding land cover is robust to data extraction approach.
441 *PLOS ONE* 14:e0216343. doi: 10.1371/journal.pone.0216343

- 442 Kuhn M (2017) caret: classification and regression training. CRAN, R Package Version 6.0-78:
443 <https://cran.r-project.org/package=caret>
- 444 Lantz SM, Gawlik DE, Cook MI (2011) The effects of water depth and emergent vegetation on
445 foraging success and habitat selection of wading birds in the Everglades. *Waterbirds*
446 34:439–447. doi: 10.1675/063.034.0406
- 447 Maechler M, Rousseeuw P, Struyf A, et al (2015) Cluster: cluster analysis basics and extensions.
448 CRAN, R Package Version 2.0.7: <https://cran.r-project.org/web/packages/cluster/index.html>
- 449 McCauley LA, Anteau MJ, van der Burg MP, Wiltermuth MT (2015) Land use and wetland
450 drainage affect water levels and dynamics of remaining wetlands. *Ecosphere* 6:art92. doi:
451 10.1890/ES14-00494.1
- 452 McCune B, Grace JB, Urban DL (2002) Analysis of ecological communities. MjM Software
453 Design, Glenden Beach, Oregon
- 454 Mensing DM, Galatowitsch SM, Tester JR (1998) Anthropogenic effects on the biodiversity of
455 riparian wetlands of a northern temperate landscape. *Journal of Environmental Management*
456 53:349–377. doi: 10.1006/jema.1998.0215
- 457 Millett B, Johnson WC, Guntenspergen G (2009) Climate trends of the North American prairie
458 pothole region 1906–2000. *Climatic Change* 93:243–267. doi: 10.1007/s10584-008-9543-5
- 459 Mitsch WJ, Gosselink JG (2015) *Wetlands*, 5th edn. Wiley, New York, NY
- 460 Oksanen J, Blanchet FG, Kindt R, et al (2017) vegan: community ecology package. R Package.
461 Version 2.4-2 1: <http://cran.r-project.org/package=vegan>
- 462 Osborn JM, Hagy HM, Mcclanahan MD, et al (2017) Habitat selection and activities of dabbling

- 463 ducks during non-breeding periods. *Journal of Wildlife Management* 81:1482–1493. doi:
464 10.1002/jwmg.21324
- 465 Peck J (2010) *Multivariate analysis for community ecologists: step-by-step using PC- ORD*.
466 MjM Software Design, Gleneden Beach, Oregon
- 467 Pöysä H (1986) Species composition and size of dabbling duck (*Anas* spp.) feeding groups: are
468 foraging interactions important determinants? *Ornis Scandinavica* 63:33–41.
- 469 R Core Team (2017) *R: A language and environment for statistical computing*. R Foundation for
470 Statistical Computing, Vienna, Austria, R Version 3.2.3: <http://www.r-project.org/>
- 471 Rashford BS, Bastian CT, Cole JG (2011) Agricultural land-use change in prairie canada:
472 implications for wetland and waterfowl habitat conservation. *Canadian Journal of*
473 *Agricultural Economics* 59:185–205. doi: 10.1111/j.1744-7976.2010.01212.x
- 474 Ripley B (2016) *Tree: classification and regression trees*. CRAN, R Package Version 1.0-37:
475 <https://cran.r-project.org/web/packages/tree/index.html>
- 476 Roberts DW (2016) *labdsv: ordination and multivariate analysis for ecology*. CRAN, R Package
477 Version 1.8-0: <https://cran.r-project.org/package=labdsv>
- 478 Rooney RC, Bayley SE, Creed IF, Wilson MJ (2012) The accuracy of land cover-based wetland
479 assessments is influenced by landscape extent. *Landscape Ecology* 27:1321–1335. doi:
480 10.1007/s10980-012-9784-6
- 481 Schindler DW, Donahue WF (2006) An impending water crisis in Canada’s western prairie
482 provinces. *Proceedings of the National Academy of Sciences* 103:7210–7216. doi:
483 10.1073/pnas.0601568103

- 484 Semlitsch RD, Bodie JR (1998) Are Small, Isolated Wetlands Expendable? Conservation
485 Biology 12:1129–1133. doi: 10.1046/j.1523-1739.1998.98166.x
- 486 Shealer DA, Alexander MJ (2013) Use of aerial imagery to assess habitat suitability and predict
487 site occupancy for a declining wetland-dependent bird. Wetlands Ecology and Management
488 21:289–296. doi: 10.1007/s11273-013-9300-z
- 489 Signorell A (2017) DescTools: tools for descriptive statistics. CRAN, R Package Version 0.99.2:
490 <https://cran.r-project.org/package=DescTools> Stewart RE, Kantrud HA (1971)
491 Classification of natural ponds and lakes in the glaciated prairie region. Washington, DC
- 492 Stoddard JL, Larsen DP, Hawkins CP, et al (2006) Setting expectations for the ecological
493 condition of streams: The concept of reference condition. Ecological Applications 16:1267–
494 1276. doi: 10.1890/1051-0761(2006)016[1267:SEFTEC]2.0.CO;2
- 495 U.S. EPA (2017) Section 404 and Swampbuster: Wetlands on Agricultural Lands.
496 <https://www.epa.gov/cwa-404/section-404-and-swampbuster-wetlands-agricultural-lands>.
497 Accessed 9 Sep 2017
- 498 Van Der Kamp G, Stolte WJ, Clark RG (1999) Drying out of small prairie wetlands after
499 conversion of their catchments from cultivation to permanent brome grass. Hydrological
500 Sciences Journal 44:387–397. doi: 10.1080/02626669909492234
- 501 Verhoeven JTA, Setter TL (2010) Agricultural use of wetlands: opportunities and limitations.
502 Annals of Botany 105:155–163. doi: 10.1093/aob/mcp172
- 503 Wilson MJ, Bayley SE (2012) Use of single versus multiple biotic communities as indicators of
504 biological integrity in northern prairie wetlands. Ecological Indicators 20:187–195. doi:

505 10.1016/j.ecolind.2012.02.009

506 Zedler JB, Kercher S (2005) Wetland resources: status, trends, ecosystem services, and

507 restorability. *Annual Review of Environment and Resources* 30:39–74. doi:

508 10.1146/annurev.energy.30.050504.144248

509 Zimmer KD, Hanson MA, Butler MG (2000) Factors influencing invertebrate communities in

510 prairie wetlands: a multivariate approach. *Canadian Journal of Fisheries and Aquatic*

511 *Sciences* 57:76–85. doi: 10.1139/f99-180

512 Zimmer KD, Hanson MA, Butler MG, Duffy WG (2001) Size distribution of aquatic

513 invertebrates in two prairie wetlands, with and without fish, with implications for

514 community production. *Freshwater Biology* 46:1373–1386. doi: 10.1046/j.1365-

515 2427.2001.00759.x

Table 1 This table provides indicator values for species belonging to each of the five bird assemblages identified via cluster analysis of the dataset including only birds categorized as wetland-dependent species. Each species is grouped under the assemblage for which it had the highest indicator value, and the table includes all 38 species, regardless of whether it was a significant indicator of an assemblage. However, only 13 species were significant indicators ($p < 0.05$), indicated by “*”. The associated p -value indicates the probability that an indicator value that large could be obtained from the data by chance alone. Note that the Hummock Nester assemblage was the first assemblage to merge (into the Pond and Reed Associates assemblage) during agglomerative clustering analysis (Figure 2) and had no significant indicators.

Group Name	Common Name	Scientific Name	Indicator Value	p -value
Shrub Associates	Song Sparrow*	<i>Melospiza melodia</i>	74.89	0.001
	Common Yellowthroat	<i>Geothlypis trichas</i>	13.33	0.169
	Gray Catbird	<i>Dumetella carolinensis</i>	10.41	0.200
	Nelson's Sparrow	<i>Ammodramus nelsoni</i>	7.28	0.499
	Lesser Scaup	<i>Aythya affinis</i>	7.54	0.602
	Long-billed Curlew	<i>Numenius americanus</i>	3.85	0.880
Wetland Edge Nesters	Marbled Godwit*	<i>Limosa fedoa</i>	37.89	0.001
	Mallard*	<i>Anas platyrhynchos</i>	35.29	0.015
	Brewer's Blackbird	<i>Euphagus cyanocephalus</i>	17.40	0.094
	Canada Goose	<i>Branta canadensis</i>	17.75	0.193
Hummock Nesters	Wilson's Snipe	<i>Gallinago delicata</i>	25.35	0.100
	Swamp Sparrow	<i>Melospiza georgiana</i>	11.76	0.308
	Bonaparte's Gull	<i>philadelphia</i>	9.62	0.337
	Lesser Yellowlegs	<i>Tringa flavipes</i>	2.44	0.980
Shorebirds	Willet*	<i>Tringa semipalmata</i>	53.45	0.001
	Wilson's Phalarope*	<i>Phalaropus tricolor</i>	74.88	0.001
	Northern Shoveler*	<i>Anas clypeata</i>	39.24	0.005
	Blue-winged Teal*	<i>Anas discors</i>	31.10	0.027
	Northern Pintail*	<i>Anas acuta</i>	26.38	0.042
	Gadwall*	<i>Anas strepera</i>	28.70	0.046
	American Avocet	<i>Recurvirostra americana</i>	19.23	0.062
	Green-winged Teal	<i>Anas crecca</i>	14.08	0.306
	American Wigeon	<i>Anas americana</i>	8.43	0.341
	Northern Rough-winged Swallow	<i>Stelgidopteryx serripennis</i>	4.36	0.606
	Horned Grebe	<i>Podiceps auritus</i>	4.12	0.818
Pond & Reed Associates	Red-winged Blackbird*	<i>Agelaius phoeniceus</i>	30.66	0.001
	Sora*	<i>Porzana carolina</i>	44.53	0.001
	American Coot*	<i>Fulica americana</i>	37.00	0.008
	Redhead*	<i>Aythya americana</i>	29.02	0.013
	Black Tern	<i>Chlidonias niger</i>	13.33	0.105
	Ring-billed Gull	<i>Larus delawarensis</i>	13.33	0.130
	Franklin's Gull	<i>Leucophaeus pipixcan</i>	11.96	0.159
Tree Swallow	<i>Tachycineta bicolor</i>	15.79	0.179	

Group Name	Common Name	Scientific Name	Indicator Value	<i>p</i> -value
	Alder Flycatcher	<i>Recurvirostra americana</i>	11.93	0.209
	Ruddy Duck	<i>Oxyura jamaicensis</i>	13.33	0.230
	Yellow-headed Blackbird	<i>Xanthocephalus xanthocephalus</i>	13.47	0.257
	Le Conte's Sparrow	<i>Ammodramus leconteii</i>	7.33	0.622
	Canvasback	<i>Aythya valisineria</i>	4.11	0.718

Table 2 Contingency table comparing the observed classification of wetland sites based on their comprehensive wetland-dependent bird assemblage and the predicted assemblage membership for each site. Predictions were based on a classification tree based on a set of local, landscape-level, and regional habitat characteristics (full list in Online Resource 2).

		Observed				
		Shrub Associates	Wetland Edge Nesters	Hummock Nesters	Shorebirds	Pond & Reed Associates
Prediction	Shrub Associates	7	0	2	0	0
	Wetland Edge Nesters	0	4	1	0	0
	Hummock Nesters	3	1	11	0	0
	Shorebirds	1	1	1	13	3
	Pond & Reed Associates	2	0	2	0	12

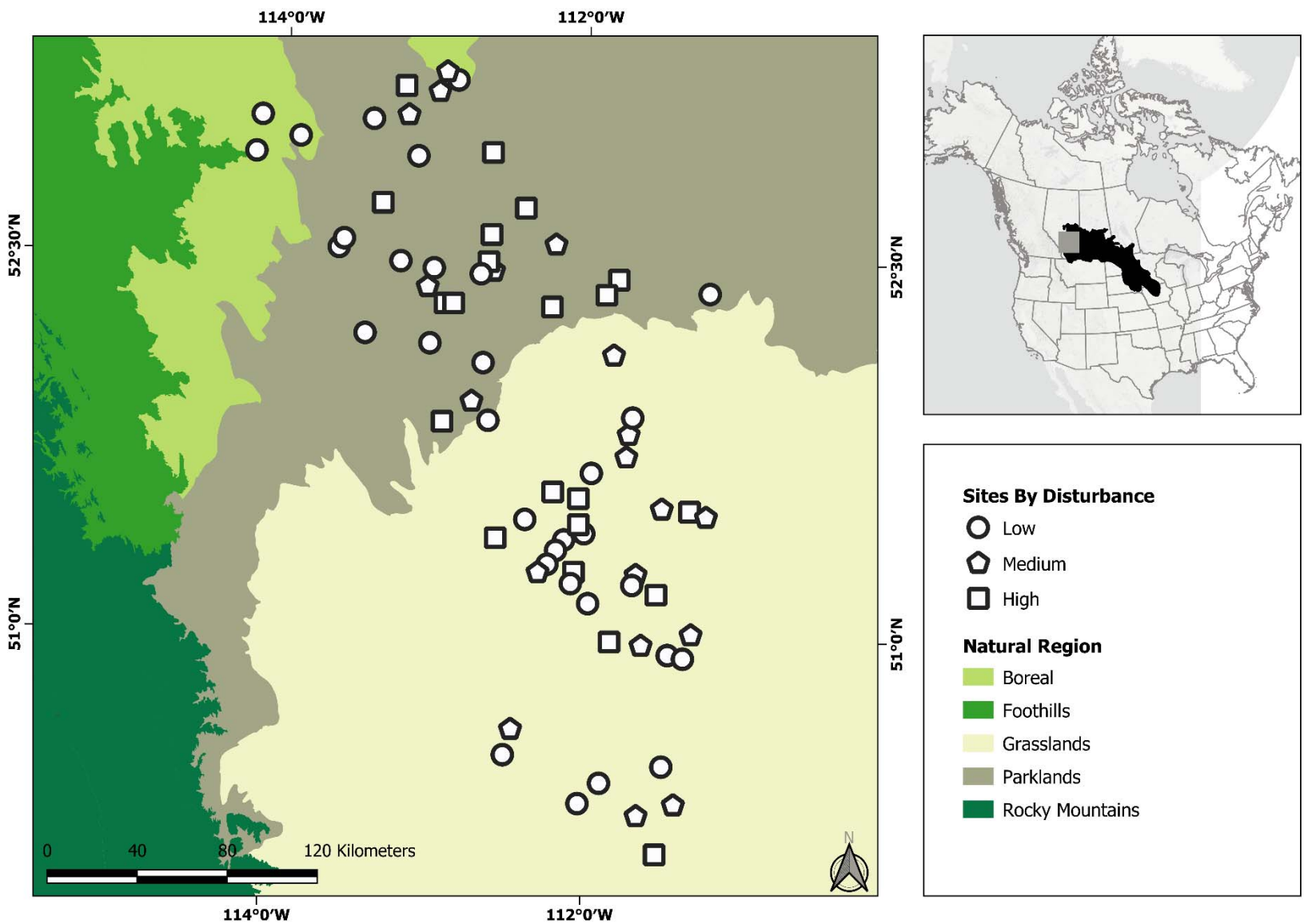
Fig. 1 A map of study our region in the northern prairie pothole region. Our 72 wetland sites occupied both the Grasslands and Parklands region, belonging to temporary ($n=11$), seasonal ($n=18$), semi-permanent ($n=10$), and permanence ($n=9$) permanence classes.

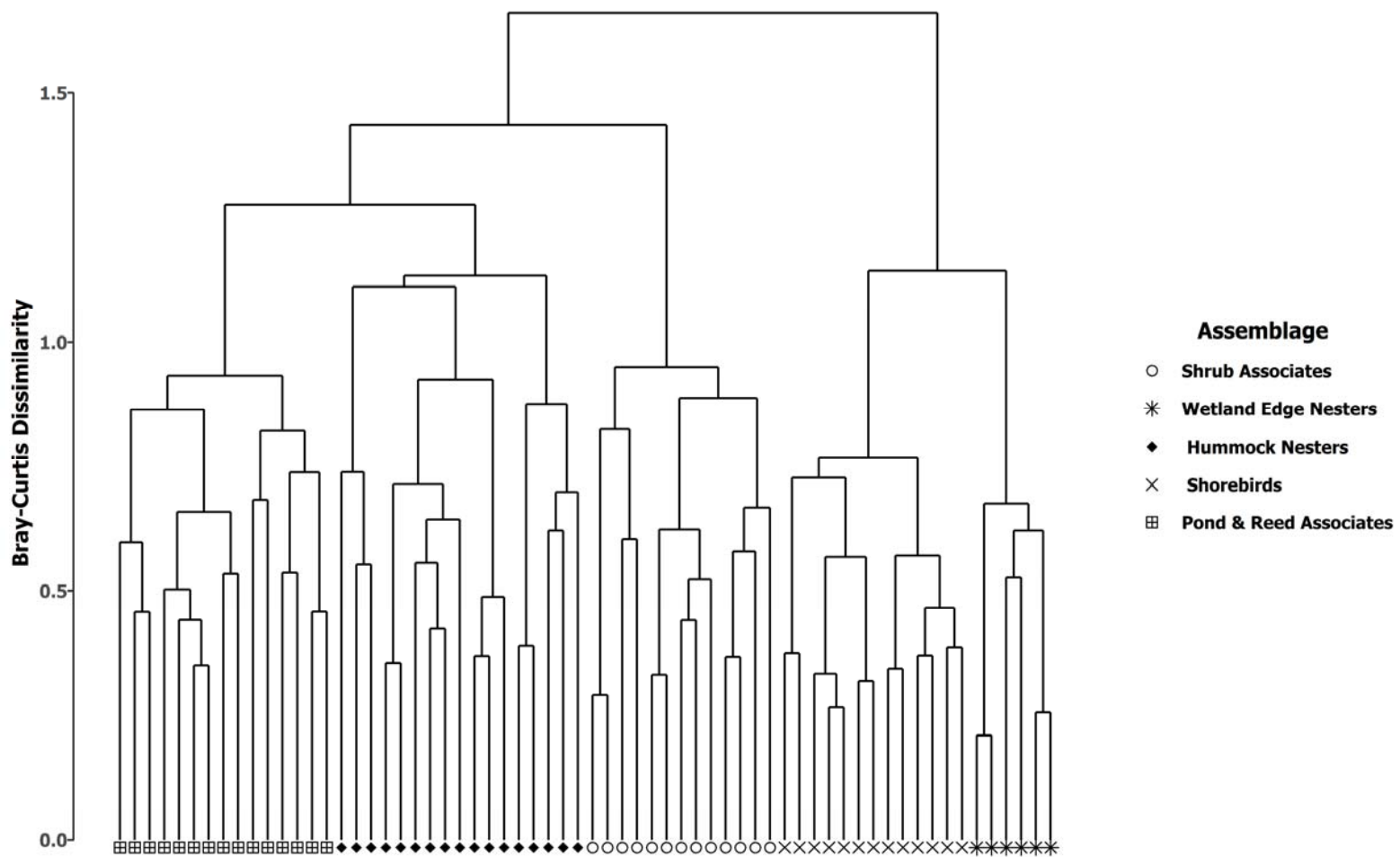
Fig. 2 Dendrogram from agglomerative hierarchical clustering, where group membership was derived from indicator species analysis for birds categorized as wetland-dependent species only. Birds categorized as terrestrial species were excluded from this analysis (see Online Resource 1). Symbology of the sites at the tips of the dendrogram reflects the optimal dendrogram pruning level, determined using indicator species analysis. Group names were based on the life history traits of the dominant indicator taxa for each group.

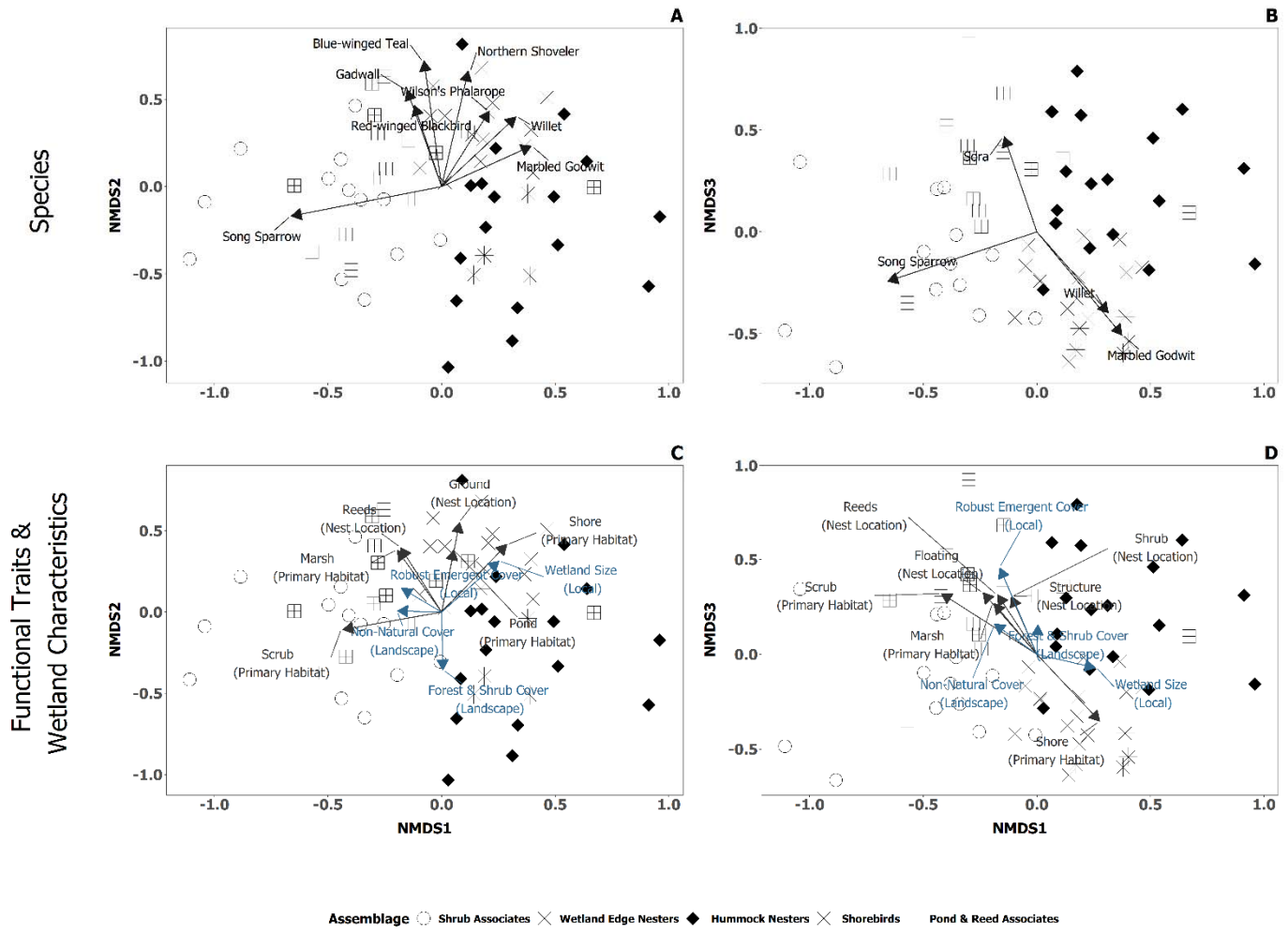
Fig. 3 Plot of the nonmetric multidimensional scaling ordination for wetland-dependent birds, for both axis one and two (A & C) and two and three (B & D). Site assemblages are the result of our paired agglomerative hierarchical clustering and indicator species analyses. We estimated vectors from correlations between both axis and the 1) abundance of wetland birds (A & B), 2) abundance of wetland birds by functional traits (A & B), 3) percentage of non-natural cover and forest and shrub cover in the landscape (C& D) and 4) wetland size and percentage cover of robust emergent vegetation in a given wetland (C & D). Species vectors shown had r^2 values greater than 0.2 for both axes; for functional traits, r^2 values were greater than 0.1. Functional trait vectors indicate a birds' nesting location and primary habitat (black), while wetland characteristics vectors indicate both local and landscape wetland characteristics (blue).

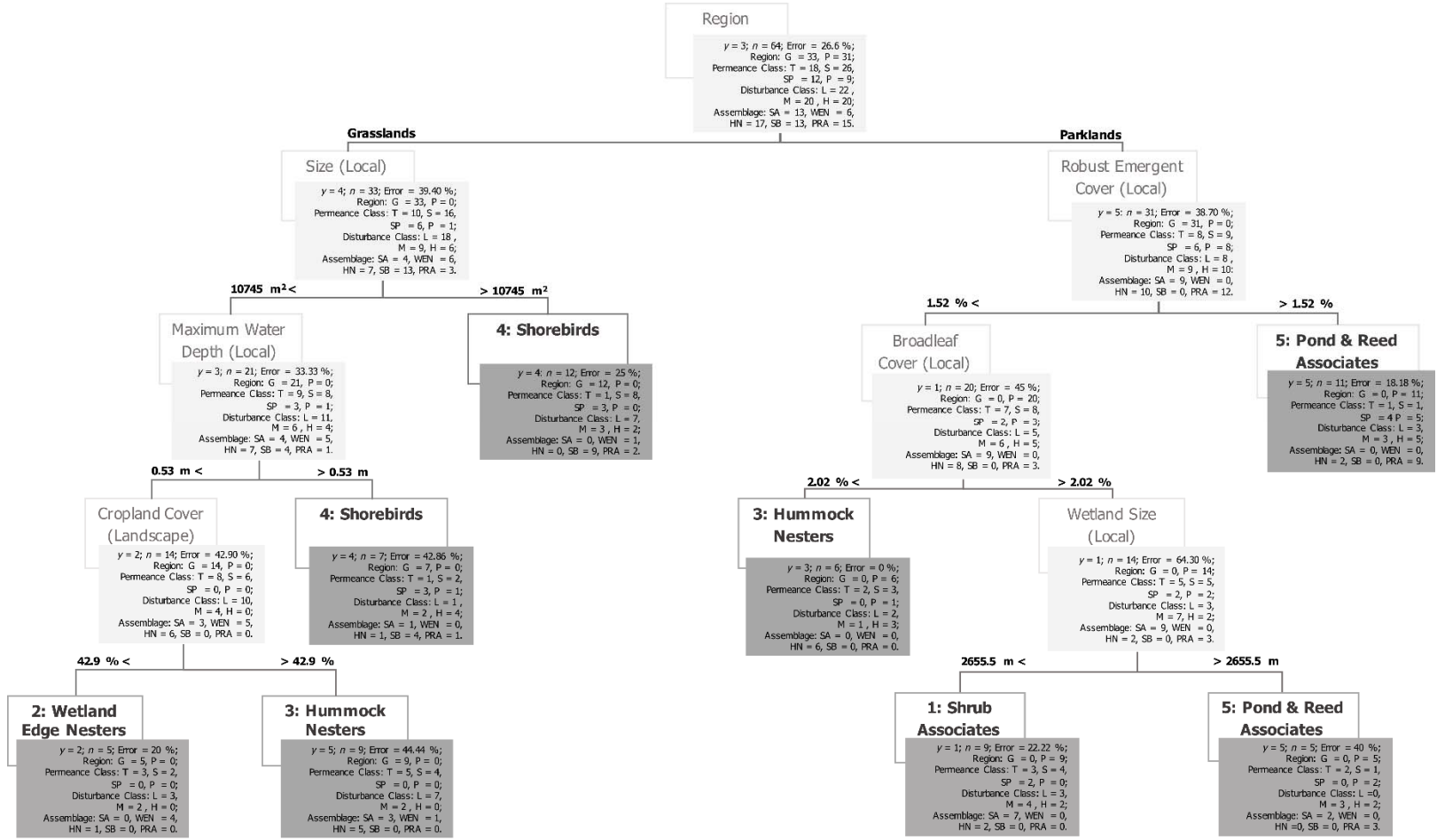
Fig. 4 Classification and regression tree using a combination of local, landscape-level, and regional variables to predict a bird's assemblage type. Assemblages are derived from cluster analysis carried out on the abundance of wetland birds we observed using the wetlands. Terminal nodes indicate the assemblage predicted to occur at this subset of wetlands by our classification

tree. For each node, we present the 1) predicted assemblage; 2) misclassification error rate; and 3) the number of sites per region (G – Grassland and P – Parkland), permanence class (T – Temporary, S – Seasonal, SP – Semi-Permanent and P – Permanent), disturbance class (L – Low: >25 %, M – Medium: <25<75 % and H – High: <75 % non-natural cover), and observed assemblage (SA: Shrub Associates, WEN: Wetland Edge Nesters, HN: Hummock Nesters, SB: Shorebirds and PRA: Pond & Reed Associates).









Online Resource 1 List of species included in our analysis. Species listed as wetland obligate – wet are known to nest in wetlands (i.e., shorebirds and waterfowl), while species listed as wetland obligate – dry (e.g., Bonaparte's Gull, Brewer's Blackbird) may nest in forests or grasslands at the wetland edge.

Common Name	Scientific Name	Wetland-Dependent Species	
		Wetland Obligate – Wet	Wetland Obligate – Dry
Alder Flycatcher	<i>Recurvirostra americana</i>		*
American Avocet	<i>Recurvirostra americana</i>	*	
American Coot	<i>Fulica americana</i>	*	
American Wigeon	<i>Anas americana</i>	*	
Baltimore Oriole	<i>Icterus galbula</i>		*
Barn Swallow	<i>Hirundo rustica</i>		*
Black-billed Magpie	<i>Pica hudsonia</i>		*
Black Tern	<i>Chlidonias niger</i>	*	
Bonaparte's Gull	<i>Chroicocephalus philadelphia</i>		*
Brewer's Blackbird	<i>Euphagus cyanocephalus</i>		*
Blue-winged Teal	<i>Anas discors</i>	*	
Canada Goose	<i>Branta canadensis</i>	*	
Canvasback	<i>Aythya valisineria</i>	*	
Clay-colored Sparrow	<i>Spizella pallida</i>		*
Common Yellowthroat	<i>Geothlypis trichas</i>		*
Eastern Kingbird	<i>Tyrannus tyrannus</i>		*
Franklin's Gull	<i>Leucophaeus pipixcan</i>	*	
Gadwall	<i>Anas strepera</i>	*	
Gray Catbird	<i>Dumetella carolinensis</i>		*
Green-winged Teal	<i>Anas crecca</i>	*	
Horned Grebe	<i>Podiceps auritus</i>	*	
Long-billed Curlew	<i>Numenius americanus</i>		*
Le Conte's Sparrow	<i>Ammodramus leconteii</i>		*
Lesser Scaup	<i>Aythya affinis</i>	*	
Lesser Yellowlegs	<i>Tringa flavipes</i>	*	
Marbled Godwit	<i>Limosa fedoa</i>		*

Common Name	Scientific Name	Wetland-Dependent Species	
		Wetland Obligate – Wet	Wetland Obligate – Dry
Mallard	<i>Anas platyrhynchos</i>	*	
Nelson's Sparrow	<i>Ammodramus nelsoni</i>	*	
Northern Pintail	<i>Anas acuta</i>	*	
Northern Rough-winged Swallow	<i>Stelgidopteryx serripennis</i>		*
Northern Shoveler	<i>Anas clypeata</i>	*	
Ring-billed Gull	<i>Larus delawarensis</i>	*	
Redhead	<i>Aythya americana</i>	*	
Red-eyed Vireo	<i>Vireo olivaceus</i>		*
Ruddy Duck	<i>Oxyura jamaicensis</i>	*	
Red-winged Blackbird	<i>Agelaius phoeniceus</i>		*
Savannah Sparrow	<i>Passerculus sandwichensis</i>		*
Sora	<i>Porzana carolina</i>	*	
Song Sparrow	<i>Melospiza melodia</i>		*
Swamp Sparrow	<i>Melospiza georgiana</i>	*	
Tree Swallow	<i>Tachycineta bicolor</i>		*
Willet	<i>Tringa semipalmata</i>	*	
Wilson's Phalarope	<i>Phalaropus tricolor</i>	*	
Wilson's Snipe	<i>Gallinago delicata</i>	*	
Yellow-headed Blackbird	<i>Xanthocephalus xanthocephalus</i>	*	

Online Resource 2 A summary of site, landscape-level and local characteristics of our study wetlands, based on the assemblages' classification from the indicator species analysis for birds categorized as wetland-dependent species. Besides region, we only used local and landscape characteristics in the classification tree. When the variable was continuous, we present the mean value with standard errors. Otherwise, we present the number of sites belonging to the category. Variables with the "*" symbol are percentage cover estimates, either at the local-level, or within a 500-m buffer landscape surrounding the wetland.

Assemblage	Shrub Associates	Wetland Edge Nesters	Hummock Nesters	Shorebirds	Pond & Reed Associates	
Site Characteristics						
<i>n</i>	13	6	17	13	15	
Permanence Class	Temporary	4	4	6	2	2
	Seasonal	6	2	7	8	2
	Semi-Permanent	2	0	3	3	4
	Permanent	1	0	1	0	7
Region	Grassland	4	6	7	13	3
	Parkland	9	0	10	0	12
Disturbance Class	Low	5	3	7	7	2
	Medium	6	2	4	3	5
	High	2	1	6	3	8
Local Characteristics						
Maximum Water Depth (m)	0.44 ± 0.06	0.26 ± 0.05	0.42 ± 0.06	0.47 ± 0.05	0.62 ± 0.07	
Wetland Size (m²)	3159.54 ± 961.17	4679.83 ± 2098.36	6030.29 ± 1907.40	14757.77 ± 2693.90	9292.07 ± 2340.74	
*Broadleaf Plants (%)	0.95 ± 0.39	5.22 ± 5.07	7.56 ± 3.15	3.28 ± 1.34	0.65 ± 0.25	
*Woody Plants (%)	0 ± 0	0 ± 0	0 ± 0	0 ± 0	0.04 ± 0.04	
*Robust Emergent Plants (%)	0.62 ± 0.57	0.22 ± 0.22	1.56 ± 1.00	0.79 ± 0.70	8.82 ± 2.76	
Landscape Characteristics						
*Non-Natural (%)	38.64 ± 9.04	34.13 ± 15.46	41.13 ± 9.42	39.4 ± 11.38	66.23 ± 7.15	
*Cropland (%)	12.46 ± 5.63	53.26 ± 15.35	30.47 ± 9.49	20.09 ± 7.83	42.48 ± 9.31	
*Urban Exposed Lands (%)	1.90 ± 0.55	3.91 ± 1.25	1.69 ± 0.48	4.01 ± 1.79	3.22 ± 0.66	
*Pasture & Forage (%)	18.09 ± 5.44	3.67 ± 1.82	4.64 ± 1.42	10.17 ± 4.05	19.5 ± 5.57	
*Water & Wetland (%)	5.64 ± 1.73	5.32 ± 2.83	7.24 ± 1.96	5.07 ± 2.18	10.87 ± 2.29	
*Forest & Shrub (%)	19.89 ± 8.89	15.87 ± 8.93	22.34 ± 7.21	2.18 ± 1.85	6.89 ± 2.47	

Online Resource 3 Cross-validation error for the classification tree for birds categorized as wetland-dependent species, based on the number of terminal nodes. We found the mean and standard error for cross-validation across 100 iterations.

Terminal Nodes	Error
10	59.74 ± 0.49
8	58.66 ± 0.47
7	60.63 ± 0.51
3	64.05 ± 0.43
2	76.25 ± 0.44
1	82.42 ± 0.30