

1 Predicting high pathogenicity avian influenza H5N1 susceptibility in wild birds

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8

9 Abstract

10 High pathogenicity avian influenza (HPAI) has caused widespread sickness and
11 mortality in wildlife, especially since the emergence of a novel H5 virus belonging to
12 clade 2.3.4.4b in 2021. The ongoing panzootic caused by this lineage has infected an
13 unprecedented diversity of species across the globe, seeming capable of impacting all
14 birds. Here, we analyse ecological and phylogenetic patterns in outbreak notifications
15 of HPAI, and predict host susceptibility to HPAI disease for Australia as the only
16 continent thus far unaffected by this panzootic. We found a significant family-level
17 phylogenetic signal, showcasing that the panzootic is not impacting all birds equally,
18 but ecological traits did not improve predictive power. Using the family-level phylogeny,
19 we predict that families of Australian seabirds, shorebirds, and waterbirds will be most
20 susceptible to HPAI once it arrives on the continent. Our results provide an empirical
21 indication of species susceptible to HPAI H5N1, which can be used to direct monitoring
22 efforts and disease management globally. With special reference to Australia, our
23 predictions can be used alongside conservation status and species-specific
24 information to inform preparedness activities, monitoring, and response upon
25 incursion.

26

27 Keywords

28 Avian influenza; HPAI H5N1; host susceptibility

29 **Introduction**

30 High pathogenicity avian influenza (HPAI) has caused problems for poultry and wildlife
31 for decades, causing significant financial and conservation harm. Low pathogenicity
32 avian influenza (LPAI) viruses are often associated with wild waterfowl, and particularly
33 ducks, and have occasionally evolved into HPAI viruses following spill over events into
34 poultry¹. HPAI has particularly surged into focus since 2021 due to the emergence of a
35 H5N1 virus belonging to clade 2.3.4.4b that is referred to as HPAI H5N1^{2,3}. The current
36 panzootic caused by this HPAI H5N1 virus is unprecedented in scale, having spread to
37 every region except Oceania (including Australia and New Zealand). The virus has
38 caused large scale mortalities in poultry and wild birds, and has increasingly also
39 spilled over into mammalian wildlife and livestock³⁻⁵. HPAI H5N1 has led to mass
40 mortality events in wildlife and cause for conservation concern in some impacted
41 species. For example, HPAI H5N1 is associated with 60% reductions in both northern
42 gannets in the UK⁶ and Dalmatian pelicans in Greece⁷, and a 91% mortality rate of
43 Caspian terns in Kazakhstan⁸. At the same time, the spread of the virus is also
44 increasingly facilitated by some of these wild bird species. The HPAI H5N1 panzootic is
45 set apart by increased host promiscuity, no longer being highly adapted specifically to
46 poultry (e.g.⁹) and spreading geographically via far-ranging waterfowl, seabirds, and
47 potentially other wild bird species¹⁰⁻¹³. As such, understanding the new disease
48 landscape for this virus, and notably what species are vulnerable to infection and may
49 play a role in the maintenance and dispersal of the virus is of considerable importance,
50 both to understand why HPAI H5N1 has had such drastic impacts on diverse wildlife
51 and to be able to sketch this panzootic's future trajectory.

52

53 HPAI H5N1 has now been detected in over 400 different avian species during the
54 current panzootic^{5,14}. Presence of LPAI, from which HPAI evolves, has a strong
55 phylogenetic signal in wild birds¹⁵, meaning avian influenza is more prevalent in certain
56 closely-related clusters of species. Notably, there is phylogenetic signal of LPAI across
57 different orders (with major reservoirs in waterfowl [Anseriformes], followed by
58 shorebirds [Charadriiformes]), but with distinct variation remaining across families and
59 species within orders¹⁵. However, how well that phylogenetic signal is preserved in the
60 current panzootic is not well understood, and is potentially very different given the

61 diversity of birds currently impacted. The apparent expansion of hosts from previous
62 clades to the current HPAI H5N1 clade 2.3.4.4b impacts the predictability of its
63 epidemiology, and notably our understanding of which species may be severely
64 impacted by HPAI H5N1 as it spreads across the world.

65

66 Transmission of HPAI H5N1 within and between species might also depend on an
67 individual's contact with the transmission pathway, i.e. how exposed they are to the
68 virus. Avian influenza transmission occurs through respiratory tracts and faecal-oral
69 pathways, which can take place directly through interaction with faecal matter or
70 indirectly through interaction with contaminated water, where the virus can persist for a
71 long time^{16,17}. Colony breeding, and specific colony traits such as distance between
72 nests, have also been implicated in HPAI H5N1 spread^{16,18}. Based on infection patterns
73 in predatory birds and mammals, HPAI H5N1 is also capable of spreading via
74 consumption of infected birds^{1,19} and potentially via kleptoparasitism²⁰. These distinct
75 transmission pathways suggest specific ecologies that increase the likelihood of birds
76 encountering the virus; association with water, likely contact with faecal matter, dense
77 flocking behaviour, and scavenging or predation are all likely to increase the probability
78 of a species encountering HPAI H5N1²¹. However, empirical testing of these ecological
79 traits that might increase virus exposure across known cases of H5N1 are generally
80 restricted to certain regions (e.g.^{18,21,22}), and thus their generality is poorly understood.
81 Improved understanding of how ecological traits can increase disease exposure will
82 furthermore improve our predictive power of which species are likely to be impacted
83 once (or 'if') HPAI H5N1 reaches the last region it hasn't infected, Oceania (including
84 Australia and New Zealand), and other more isolated parts of the world that have so far
85 escaped exposure to the virus.

86

87 In this paper, we evaluate the influence of ecological traits and phylogenetic
88 relationships on species' HPAI notifications and use this to predict susceptibility to
89 HPAI disease of naïve, Australian species. We analysed notifications of HPAI in wild
90 birds reported to the World Organisation of Animal Health (WOAH) across the world
91 since the start of the panzootic as our indicator of susceptibility to disease. Using
92 phylogenetic generalised linear mixed models, we modelled notifications of HPAI H5N1

93 in wild birds against multiple predictors: a family-level phylogeny and ecological traits
94 that might influence disease exposure (habitat, diet, and congregation behaviour).
95 Following model selection, we predict HPAI H5N1 disease susceptibility in Australian
96 birds, who remain unexposed to the current panzootic. Here, our measure of
97 “susceptibility” is the predicted number of HPAI notifications, which is modelled based
98 on HPAI notifications to WOAAH.

99

100 **Methods**

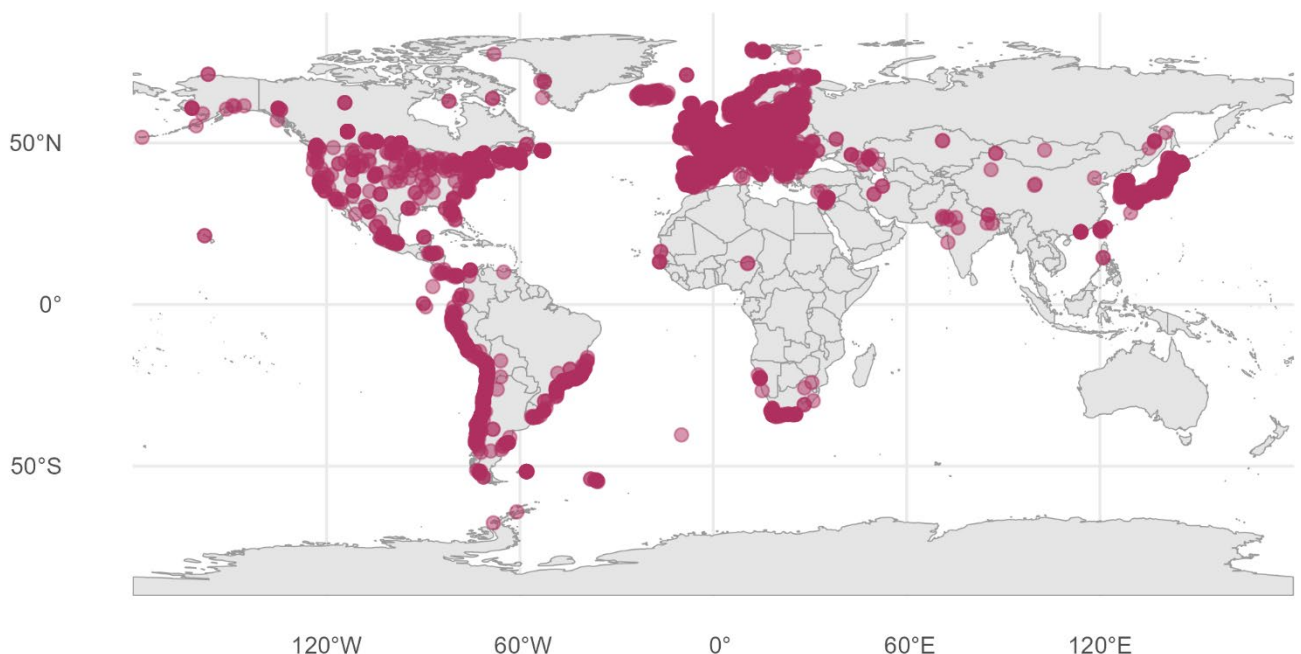
101 To model factors predicting HPAI H5N1 notifications in wild birds, we used the WOAAH
102 World Animal Health Information System (WAHIS) database of HPAI notifications.
103 WOAAH member countries are expected to submit information on a range of animal
104 diseases, and for each disease notification must provide information on the number of
105 outbreaks and cases, as well as biosecurity measures taken²³. An HPAI notification in
106 the WAHIS database can represent a) a notification of an HPAI detection in an
107 environmental sample from the recorded wild bird species, b) a notification of an HPAI
108 detection in the species, but where the individual had no obvious or reported clinical
109 signs of sickness/death, and c) a notification of HPAI in sick and dead wild birds of the
110 reported species. Importantly, each notification may represent a single bird or multiple
111 birds from a single species. Therefore, by using ‘notifications’ rather than ‘number of
112 impacted birds’ as our input data, we avoid overrepresentations from extreme mass
113 mortality events. While WOAAH members are expected to submit notifications on HPAI
114 outbreaks, there is likely a certain amount of underreporting of HPAI to the database⁵
115 which may lead to certain biases. For example, HPAI notifications to WAHIS correlate
116 with human density⁵, since diseased birds are more easily observed where there are
117 people. Similarly, it has been found that large-bodied species are more likely to be
118 detected with HPAI because they are easier to observe than small species²².
119 Furthermore, HPAI notifications are likely to largely reflect instances of diseased birds,
120 meaning infected but apparently healthy birds are likely to be underreported. While
121 there is likely underreporting that has important implications for the biases inherent in
122 the dataset, the WAHIS database still provides a minimum indication of HPAI
123 notifications (and may therefore underestimate, rather than overestimate, HPAI events).
124 The database was accessed on 16/02/2026 and filtered to only include notifications of

125 outbreaks reported in wild birds of known species since October 2021. In 99% of the
126 cases, the subtype of the HPAI notification was evaluated and established to be HPAI
127 H5. We therefore assume that the majority (if not all) notifications used in our study
128 relate to HPAI H5N1. It should be noted that WOAH bears no responsibility for the
129 integrity or accuracy of the data, including but not limited to any deletion, manipulation,
130 or reformatting of data that may have occurred beyond its control.

131

132 HPAI H5N1 has spread through different world regions since 2021, with more severe
133 impacts in different regions at different times. To account for the spatial spread of HPAI
134 H5N1 across large regions, we grouped HPAI notifications by region. Region was
135 determined using the 'countrycode' package²⁴ in R, where each country was grouped to
136 one of seven regions: 'East Asia and Pacific', 'South Asia', 'Middle East and North Africa',
137 'Sub-Saharan Africa', 'Europe and Central Asia', 'North America', 'Latin America and
138 Caribbean', and 'Antarctica'. In the resulting dataset, we had number of HPAI
139 notifications per species per region. The full spatial distribution of HPAI H5N1
140 notifications in wild birds made to WAHIS in our dataset is show in Figure 1.

141



142

143 *Figure 1: Map of WAHIS HPAI H5N1 notifications in wild birds. Each dot represents the*
144 *reported location of a HPAI H5N1 notification, and can represent one or several birds of*
145 *a species. There is likely underreporting of HPAI notifications to WAHIS, but the data*

146 *represents the minimum HPAI H5N1 notifications. The data was accessed on*
147 *16/02/2026 and filtered to include notifications after October 2021.*

148

149 Because of the links between disease transmission pathways and ecological traits like
150 aquatic lifestyles, certain feeding ecologies, and tendency to congregate, we modelled
151 how ecological traits might influence HPAI H5N1 notifications in wild birds. Initial
152 ecological categorisations of habitat and trophic niche were obtained from Avonet²⁵. In
153 our analyses, we wanted to avoid categorisations that were too narrow (e.g.
154 differentiating between frugivores and granivores) or perhaps arbitrarily differentiated
155 between species with otherwise similar ecologies (e.g. denominating the Common
156 Merganser as inhabiting “riverine” habitat, but other mergansers as “wetland”).
157 Therefore, we modified some habitat and trophic niche categorisations based on
158 information in Birds of the World²⁶, and broadened the groupings. We thus had three
159 categories for habitat: Terrestrial, Freshwater, and Coastal/Marine, and four categories
160 for diet: Predators (including vertebrate predators and scavengers), Piscivores, Plant-
161 based diets (including aquatic and terrestrial plant material), and Omnivores (including
162 any species that were both predators and plant-based feeders, and invertebrate
163 predators as these often also take plant matter). We used BirdLife’s list of congregating
164 birds as our initial starting points for whether species were known to congregate or not
165 (Y/N), and supplemented this with information from Birds of the World²⁶. For a full list of
166 species and our ultimate ecological categorisation for these species, see Supporting
167 Information Table S1.

168

169 To predict HPAI disease susceptibility in Australian birds in the event of HPAI H5N1
170 incursion into Australia, we used the BirdLife Working List of Australian Birds dataset
171 (<https://birddata.birdlife.org.au/whats-in-a-name>) to generate a list of Australian bird
172 species. The list was refined to exclude rare vagrants and uncommon non-native
173 species. Similar to how we treated the WAHIS dataset, we used Avonet’s and BirdLife’s
174 ecological data on habitat, diet, and congregation as starting points, with refinement
175 and broadening of categories to generate matching ecological traits. The full list of
176 Australian birds we used, their ecological categorisations, and their IUCN status can be
177 found in Supporting Information Table S2.

178

179 *Statistical analyses*

180 All analyses were conducted in R version 4.4.0²⁷.

181

182 To model HPAI H5N1 notifications in birds, we used a phylogenetic generalised linear
183 mixed model (GLMM) in the ‘brms’ package²⁸. In all the following models run in ‘brms’,
184 we used the flat default priors and ran models for 5000 iterations, with 600 iteration
185 warmup and a thinning interval of 5. Model convergence was assessed by visual
186 inspection of chains and by posteriors predictive checks of the models (using the
187 ‘pp_check’ function). We used a Poisson distribution of the number of HPAI
188 notifications per species per region (region defined using the ‘countrycode’ package as
189 outlined above), controlling for region as a fixed effect and running the model under a
190 phylogenetic framework. For the phylogeny, we used the family-level phylogeny from
191 Kuhl et al²⁹. We used family, rather than species-level phylogeny to avoid reporting
192 biases for more common species, when species in the same family are likely to share
193 similar ecological traits and immune system architecture. This was especially relevant
194 for our next step, outlined below, wherein we used the model of HPAI notifications to
195 predict HPAI H5N1 disease susceptibility in Australian species (we wanted to avoid
196 drastically uneven outbreak notification estimates for Australian species in the same
197 family, but where some species were closely related to a species with high HPAI
198 notifications). We built upon the phylogenetic GLMM to include the ecological traits of
199 species: habitat (N = 3 categories), diet (N = 4 categories), and whether the species is
200 known to congregate (Y/N). The models did not include any offsets. We fitted three
201 models that had one ecological predictor (habitat, diet, or congregation) in addition to
202 region as fixed effects, and then an additional model that included all 3 ecological
203 predictors and region. We evaluated model fit of these against the null, region and
204 phylogeny-only model using leave-one-out (LOO) cross validation information criterion
205 (IC), which is interpreted similarly to AIC where low values are associated with better
206 models.

207

208 The next step in our analyses was to predict which Australian species may be
209 susceptible to HPAI H5N1 once (or ‘if’) it arrives in Australia, based on patterns of HPAI

210 H5N1 notifications elsewhere in the world. We used the HPAI notification data to
211 predict numbers of HPAI H5N1 notifications in Australian birds, and use this as our
212 metric of predicted susceptibility to disease from HPAI H5N1. Most (~98%) of HPAI
213 notifications in the WAHIS database since October 2021 report deaths for species that
214 have outbreak notifications, which means that our predicted susceptibility is also linked
215 to a species' likelihood of experiencing sickness and death. To predict susceptibility to
216 HPAI H5N1, we first added the Australian bird families to our above phylogeny²⁹, thus
217 resulting in a phylogeny with the families in the WAHIS database and the Australian
218 families. Using the 'castor' package³⁰, we predicted HPAI H5N1 notification likelihood
219 onto the Australian species. This was done using hidden state prediction via
220 phylogenetic independent contrasts, using the family-level phylogeny with both known
221 and unknown HPAI notifications (wherein known HPAI notifications were expressed as
222 an average per family). Through this, we retrieved predicted HPAI H5N1 notifications for
223 Australian bird families, which we interpreted as their predicted susceptibility to
224 disease from HPAI H5N1.

225

226 Plots were made using 'ggtree'³¹ and 'ggplot2'³². Lastly, we extracted IUCN Redlist status
227 for Australian species using the 'rredlist' package³³, to ascertain the conservation status
228 of any species predicted to be highly susceptible to HPAI H5N1 and thus highlight
229 species that may be at greater risk due to pre-existing vulnerabilities or due to other
230 reasons than vulnerability to HPAI H5N1.

231

232 *Animal ethics statement*

233 Our data is sourced from the WOAHA WAHIS database of global HPAI H5N1 notifications
234 in wild birds. Therefore, our analysis is conducted on a pre-existing dataset with no new
235 data collected for the purpose of this study. We have no permit details to report.

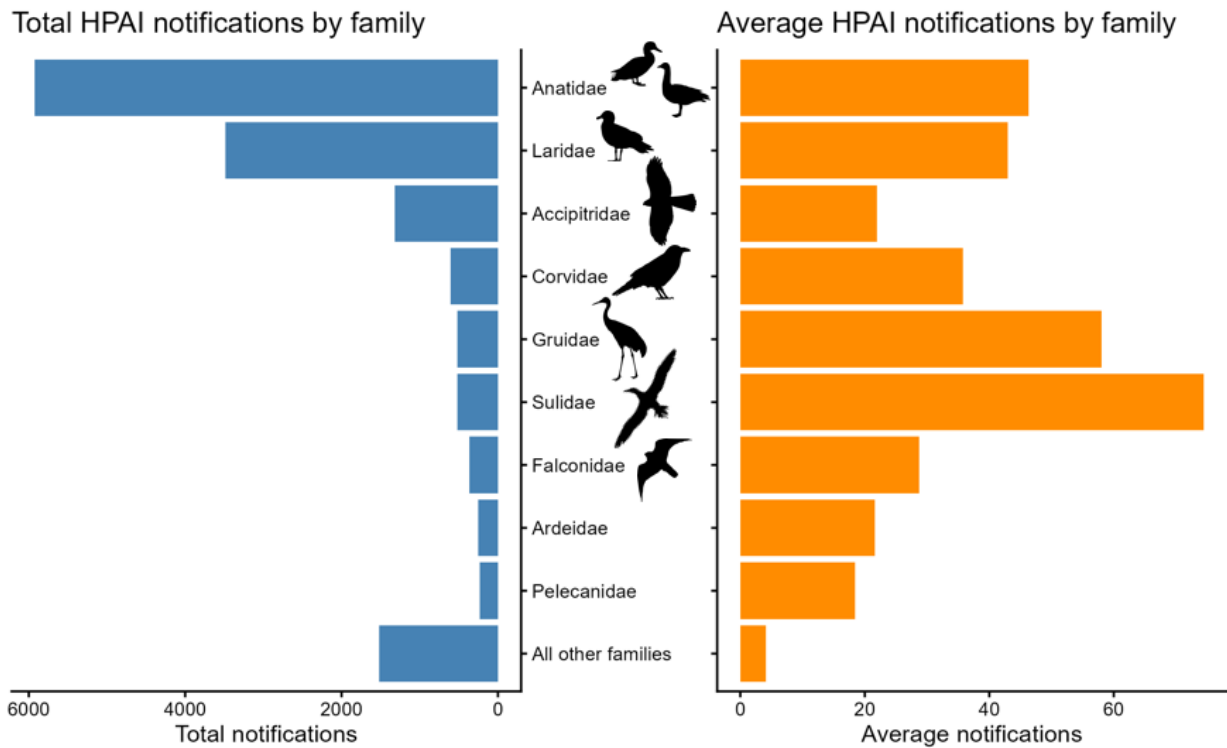
236

237 **Results**

238 When analysing notifications of HPAI H5N1 in wild birds since October 2021 using the
239 phylogeny-only model, we found a statistically significant phylogenetic signal (Pagel's λ :
240 0.56, 95% CI: 0.22 – 0.84). There were predominantly high numbers of total and average
241 HPAI H5N1 notifications amongst Anatidae (ducks, geese, and swans), Laridae (gulls,

242 terns, and noddies), Sulidae (gannets and boobies), and Gruidae (cranes). To a lesser
 243 extent, crows and ravens (Corvidae), birds of prey (like Falconidae and Accipitridae),
 244 and other seabirds (like Pelecanidae [pelicans] and Alcidae [auks]) also had higher
 245 numbers of HPAI H5N1 notifications (Figure 2).

246



247

248 *Figure 2. HPAI H5N1 notifications in wild birds 2021 – 2024. The left panel shows total*
 249 *HPAI H5N1 notifications made to WOA/WAHIS per family, while the right panel shows*
 250 *average HPAI H5N1 notifications per family. A few key families are highlighted by*
 251 *inclusion of bird icons from phylopic.org, going down from the top: Anatidae,*
 252 *Accipitridae, Corvidae, Gruidae, Sulidae, and Falconidae.*

253

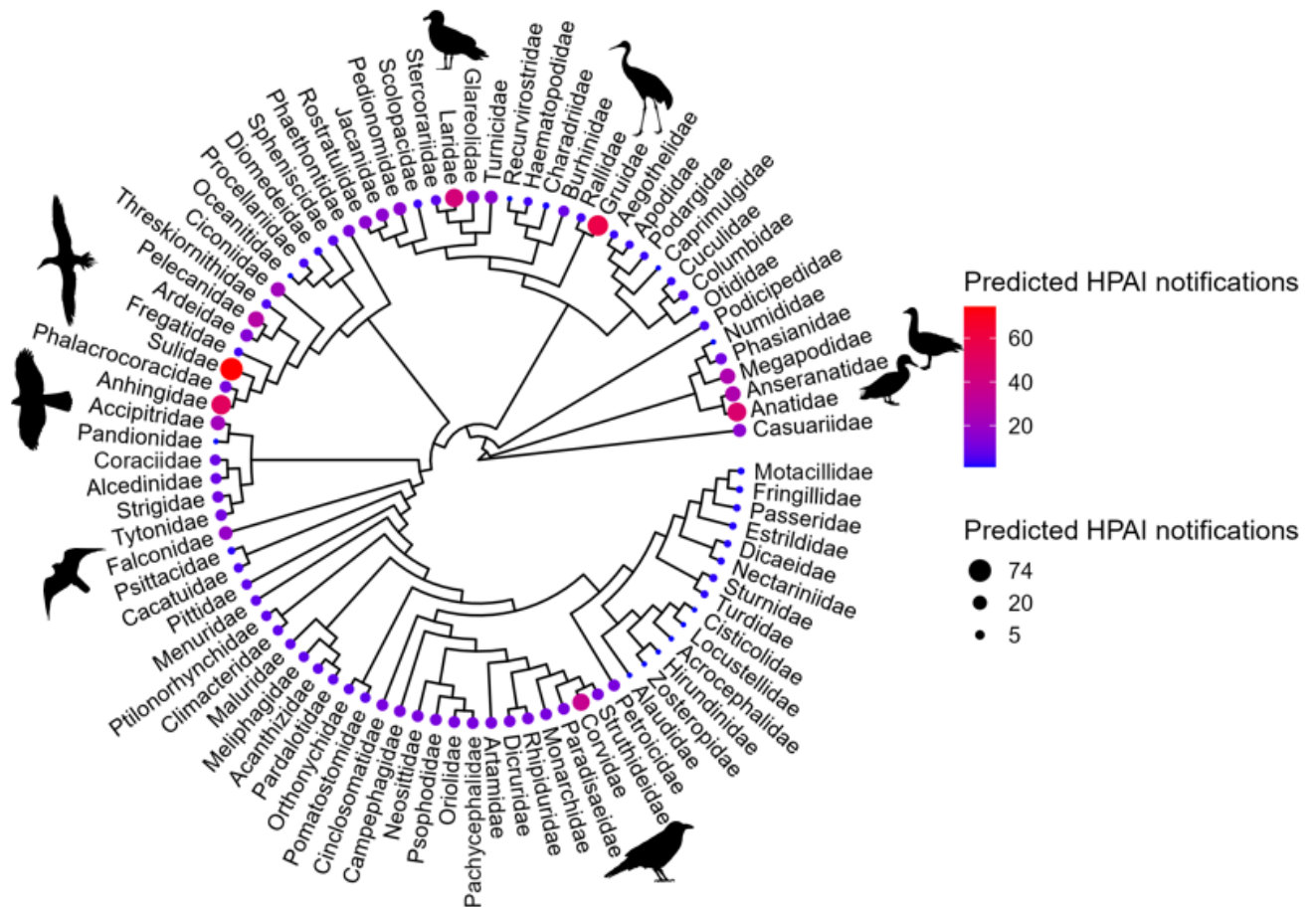
254 Using leave-one-out (LOO) cross validation, we compared the fit of this null, region and
 255 phylogeny-only model and that of the ecological traits models. There were 4 models
 256 with ecological traits, where 3 models consisted of a single ecological trait (habitat,
 257 diet, or congregation) and a fourth model that included all three ecological traits (see
 258 Table S3 for summary statistics of all models). All four of these models included the
 259 phylogeny and region (see Methods). There was substantial overlap in the standard
 260 errors of the LOO ICs computed for the models, which means that the additional

261 variables in our ecological models did not significantly improve model fit over the null,
262 phylogeny-only model (Figure S1;³⁴). Therefore, we present results for the simpler,
263 phylogeny-only model.

264

265 When predicting Australian species' HPAI H5N1 disease susceptibility (defined as their
266 predicted HPAI H5N1 notifications), the predominant groupings of predicted
267 notifications were similar to that of the training model (Figure 3). Based on global HPAI
268 H5N1 notification data since 2021 and the family-level phylogeny, the highest predicted
269 HPAI H5N1 susceptibility was predicted for Australian Sulidae (gannets and boobies),
270 followed by Gruidae (cranes), Anhingidae (darters), Anatidae (ducks, geese, and swans),
271 and Laridae (gulls, terns, and noddies). Specifically, the model predicted 74
272 notifications (i.e. HPAI detections in environmental, living, or dead bird samples) of HPAI
273 H5N1 in Sulidae family members, followed by 58 in Gruidae, 51 in Anhingidae, 46 in
274 Anatidae, and 43 in Laridae family members. Other Australian bird families, like
275 Pelecanidae (pelicans), Corvidae (crows and ravens), and Accipitridae (falcons,
276 hobbies, and kestrels), were also predicted to be susceptible to disease. Furthermore,
277 some families endemic to Australia, such as Anseranatidae (containing the magpie
278 goose – *Anseranas semipalmata*), were predicted to be moderately susceptible to HPAI
279 H5N1 with a predicted 26 notifications. Predicted HPAI H5N1 notifications for all
280 Australian bird families in our list are reported in Supporting Information Table S4.

281



282

283 *Figure 3. Predicted HPAI H5N1 susceptibility for Australian bird families. Each tip*
 284 *denotes a family, with the size and colour of the tip representing the predicted number*
 285 *of HPAI H5N1 notifications. Key families are highlighted by inclusion of bird icons from*
 286 *phylopic.org, going clockwise from the top: Laridae, Gruidae, Anatidae, Corvidae,*
 287 *Falconidae, Accipitridae, and Sulidae.*

288

289 **Discussion**

290 HPAI H5N1 2.3.4.4b has caused a panzootic of unprecedented scale⁵, but has not yet
 291 spread to Australia⁴. Here, we modelled HPAI H5N1 notifications as a function of
 292 ecology and family-level phylogeny, finding that family-level phylogeny best explains
 293 number of HPAI H5N1 notifications. The importance of host phylogeny in explaining
 294 avian influenza prevalence has been previously noted for low pathogenicity viruses¹⁵
 295 and for HPAI H5N1 in different regions (e.g.^{10,12}), which we now expand for a global
 296 dataset of HPAI H5N1. Furthermore, we use phylogeny to predict HPAI H5N1
 297 notifications in Australian birds (including for Australian endemic birds), as a metric of

298 susceptibility to HPAI H5N1 infection and resulting disease once (or 'if') it reaches the
299 continent.

300

301 While transmission of HPAI H5N1 is believed to link to ecological traits related to e.g.
302 (aquatic) habitat choice, (dabbling) foraging strategies, predation, and congregation,
303 including these ecological traits in our model did not significantly improve model fit.
304 However, we do not believe the support for our phylogeny-only model means that
305 ecological traits are not important – rather, there are specific traits as well as
306 combinations of traits yielding scenarios of probable disease transmission that are
307 likely captured by the family-level phylogeny. For example, Anatidae (ducks, geese, and
308 swans) have a high number of HPAI notifications globally and are among the families
309 predicted to be most susceptible to disease from HPAI H5N1 in Australia, likely due to
310 their aquatic lifestyle. Conversely, we hypothesized terrestrial birds to have lower HPAI
311 H5N1 notifications due to largely avoiding contact with HPAI virus contaminated water,
312 but many birds of prey (which have high HPAI H5N1 notifications) are terrestrial
313 predators. This specific interaction between diet and habitat may be important in
314 predicting HPAI H5N1 notifications, but it is captured already in the family-level
315 phylogeny, as such traits tend to be shared across members of a family and even entire
316 orders. The drawback of our approach is that the predicted HPAI H5N1 notifications are
317 generalised across species in a family. Generalising across families may be especially
318 penalizing for species that are ecological outliers compared to others within their
319 family. For example, our study predicts high HPAI H5N1 notifications for the
320 Australasian wood duck (*Chenonetta jubata*), despite its ecology differing from other
321 ducks (it is an exclusively grazing duck, while many other ducks engage in filter feeding
322 and dabbling) and its previous identification as an outlier in having low LPAI virus and
323 seroprevalence¹⁵. Similarly, subtle differences in the type of congregation behaviour can
324 seemingly drive some differences in HPAI susceptibility between closely related
325 species, such as the relatively low effect of HPAI on little terns (*Sternula albifrons*)
326 compared to other terns, which was attributed to bigger spacing between nests and
327 their tendency for single-species colonies³⁵. Despite such exceptions, our predictions
328 can serve as an initial guideline of species likely to be impacted by HPAI H5N1, with

329 additional information such as species' conservation status, population size, and a
330 variety of site- and species-specific factors used to assess potential local impacts.

331

332 In our prediction of HPAI H5N1 susceptibility in Australian birds, we define susceptibility
333 as the predicted number of HPAI H5N1 notification for a taxonomic family (where high
334 numbers of predicted HPAI H5N1 notifications is interpreted as high susceptibility to
335 disease; with 74 HPAI H5N1 notifications in Sulidae being the highest score). This
336 modelling is based on data of "outbreak notifications" from WOAHA WAHIS since
337 October 2021 (the onset of the current panzootic), where most (~98%) species with
338 notifications also have reported deaths from HPAI. This means that our predicted HPAI
339 H5N1 susceptibility reflects how easily different birds become infected and
340 subsequently die of HPAI. However, our susceptibility predictions largely ignore the role
341 of different birds in maintaining and spreading HPAI H5N1, since different species might
342 survive (and some may indeed not die from infection) and carry the virus for different
343 lengths of time. For example, bald eagles (*Haliaeetus leucocephalus*) had higher HPAI
344 H5N1 seroprevalence (indicating higher survival rate) than other birds of prey³⁶, and
345 anti-H5 antibodies in seabird eggs were higher in common eiders compared to other
346 seabird species (such as gannets, which suffered HPAI-related mass mortality
347 events³⁷). In the current HPAI panzootic, recent research has shown that the host
348 dynamics of H5N1 differs between virus genotypes³⁸, showcasing the wide range of
349 birds capable of contributing to the spread of HPAI and that the reservoir community
350 can change rapidly. Furthermore, migratory and partially-migratory birds are likely to
351 have a significant role in sustained transmission of HPAI H5N1 to other birds, as has
352 been found in other regions^{10,12,13}. This suggests that species can play different roles in
353 maintenance and spread of HPAI H5N1 after exposure to the virus, which is important
354 to consider when predicting HPAI H5N1 susceptibility. How HPAI H5N1 spread may look
355 like within the Australian continent is further complicated by the often nomadic
356 movements of Australian waterfowl as they move in response to ephemeral wetlands³⁹.

357

358 In this study, we used a family-level phylogeny to avoid biases associated with
359 particular outlier species. Our approach may still carry some inherent biases, for
360 example if a family is very speciose, very abundant, or contains very commonly

361 sampled species. However, when comparing the mean number of HPAI H5N1
362 notifications per family to the number of species in that family, the correlation was low
363 ($R^2 = -0.06$), meaning it is unlikely biases related to number of species in a family are
364 entirely driving our predictions. Indeed, because we analysed the data using the family-
365 level phylogeny that considers HPAI H5N1 notifications across a family (rather than just
366 the total), we avoid some of the exaggerated total HPAI H5N1 notifications associated
367 with very speciose and common families like Anatidae and Laridae (Figure 1). However,
368 it may also be argued that this introduces its own form of bias, if it “punishes” the HPAI
369 H5N1 susceptibility predictions for speciose and common families (hence why our
370 model predicts Sulidae, rather than Anatidae, to be the most susceptible Australian
371 family; Figure 1). It is also worth noting that our approach is inherently biased by people
372 sampling for and reporting notifications of HPAI H5N1, where real numbers of HPAI
373 H5N1 likely exceed recorded notifications by an order of magnitude⁵. By relying on
374 human reporting, there is also the possibility that the dataset we use might be biased
375 towards more frequent reporting of large birds, or similar traits that influence
376 detectability²². Furthermore, differences in sampling effort between regions may
377 exacerbate such biases, if certain families are more common in sparsely sampled
378 regions and are thus more less represented in WAHIS¹⁴.

379

380 Partly because of biases in testing and reporting HPAI H5N1 outbreaks, we did not
381 employ a presence/absence approach to modelling HPAI H5N1 notifications and
382 predicting susceptibility to disease in Australian species. Biases in testing and
383 reporting, or lack of sufficient ornithological knowledge to identify specific species,
384 mean we cannot assume that species absent from the WAHIS dataset of HPAI H5N1
385 notifications truly never had cases of HPAI H5N1, and thus we cannot assume that HPAI
386 is absent. However, our approach of using numbers of HPAI H5N1 notifications still
387 suffers from part of this bias and is likely to have influenced some of our predictions for
388 Australian species. For example, Anhingidae (darters) are amongst families with the
389 highest predicted notifications in Australia, but their non-Australian species are not
390 currently represented in the WAHIS database. However, the American darter’s (*Anhinga*
391 *anhinga*) distribution includes regions severely impacted by the current panzootic,
392 making it likely the species has encountered the virus but that it just has not been

393 detected, tested and reported to WAHIS. The family's lack of representation in WAHIS
394 means the model used the Anhingidae phylogenetic information, and its proximity to
395 Sulidae and Phalacrocoracidae (the latter of which is also underestimated in WAHIS⁴⁰),
396 to estimate a value between the two other families. The lack of Anhingidae
397 representation in the WAHIS database, despite its probable interface with the virus,
398 means the model may have over-estimated the susceptibility of Australasian darters to
399 HPAI H5N1 based on its relationship to Sulidae. An opposite scenario may also be
400 possible and potentially detrimental: in some cases, our model may have falsely
401 predicted a family as *not* susceptible to HPAI H5N1. This further highlights the
402 importance of not relying on our predicted susceptibility in isolation, but also
403 considering additional information. The wide host range of the current panzootic
404 highlights that many species are capable of contracting the virus, however we need
405 active monitoring of diverse species to confirm which families do and do not develop
406 disease. However, lack of notifications and our assumptions of relative completeness in
407 the WAHIS database impacts our predictions for Australian families susceptible to HPAI
408 H5N1.

409

410 In our analysis, adding ecological traits like diet and habitat did not significantly improve
411 the predictability of HPAI H5N1 notifications above our null, phylogeny-only model.
412 However, this does not mean those traits are not still important to consider when
413 assessing virus incursion into new ranges, like Oceania. For example, traits like colony-
414 breeding can amplify the risk to a species if the virus is able to spread rapidly through a
415 large number of birds¹⁸, but this also means there is a temporally-stratified risk to
416 consider as colony-breeding is only a relevant trait at certain times of the year²¹.
417 Australia hosts big breeding colonies of gannets, shearwaters, and other notable
418 seabirds, which might expose them to the same colony-wide mass mortalities noted
419 elsewhere^{6,40,41}. Therefore, even if dense flocking behaviour was not a major predictor of
420 HPAI H5N1 notifications in our analysis, it is a trait worth bearing in mind when
421 considering conservation impacts of potential HPAI H5N1 arrival in Australia. The same
422 holds for diet and habitat traits, which were kept deliberately broad in our analysis. For
423 example, the ecologies of geese and passerines were in this analysis occasionally
424 grouped as 'terrestrial omnivores', but their distinct feeding ecologies and habitat

425 movements can mean they will play very different roles in transmission of an HPAI H5N1
426 incursion into Australia. Similarly, analysis of the geospatial distribution of HPAI H5N1
427 has revealed that landscape characteristics can also influence risk²¹, which will be
428 crucial to consider when predicting potential incursion and spread in the Australian
429 continent.

430

431 A strength of our phylogeny-based approach to model and predict HPAI notifications is
432 that it likely captures similarity in immune architecture between closely related species,
433 in addition to the ecological similarities it captures. However, there are notable
434 exceptions to the expectation that closely related species share genomic similarities.
435 Such differences in species' immune architecture may influence their final
436 susceptibility to HPAI, and thus represents another aspect worth considering when
437 predicting HPAI H5N1 susceptibility in new ranges⁴². For example, the Australian black
438 swan (*Cygnus atratus*) is more vulnerable to HPAI than white swans and some geese⁴²,
439 likely because it lacks receptors for viral pattern recognition and has a poor immune
440 response to HPAI⁴³. These differences set black swans apart even from closely related
441 species, like the mute swan (*Cygnus olor*). Should similar deficiencies in immune
442 system architecture exist for other Australian birds, it is possible the HPAI H5N1
443 susceptibility for some Australian birds is underestimated in our analysis. While our
444 predictions of Australian species' susceptibility to disease from HPAI H5N1 can
445 function as an important indicator of what is to come, expanded genomic and
446 transcriptomic testing can further fine-tune such predictions.

447

448 Among other factors that can be important to consider when predicting HPAI H5N1
449 susceptibility in Australian birds is conservation status, where rampant disease spread
450 may have a larger impact on more vulnerable populations²². In Australian species with
451 highest predicted HPAI H5N1 notifications (within the top 50 predicted HPAI H5N1
452 notifications), all but three are listed as Least Concern on the IUCN Red List status.
453 Only the fairy tern (*Sternula nereis*), sooty tern (*Onychoprion fuscatus*), and the sarus
454 crane (*Grus antigone*) are listed as Vulnerable. However, expanding this to species
455 predicted to be moderately susceptible to HPAI H5N1 (top 80 predicted disease
456 notifications), there is one species listed as Endangered (red goshawk – *Erythrotriorchis*

457 *radiatus*), two additional birds listed as Vulnerable (grey falcon – *Falco hypoleucos* and
458 malleefowl – *Leipoa ocellata*), and one Near Threatened (letter-winged kite – *Elanus*
459 *scriptus*). It is notable that most of these are birds of prey, suggesting that while HPAI
460 H5N1 is primarily predicted to impact waterbirds, the notable conservation impacts of
461 potential HPAI H5N1 incursion into Australia may focus on predators. The impact to
462 predators might be similar to effects seen elsewhere²¹, such as declining peregrine
463 falcon (*Falco peregrinus*) populations in the Netherlands, where over 80% of tested
464 dead birds were infected with HPAI H5N1⁴⁴. Conservation vulnerability of predators to
465 HPAI H5N1 also underscores the potential conservation concerns to mammalian
466 predators¹⁹, as has been noted in South American pinnipeds⁴⁵.

467

468 **Conclusion**

469 HPAI H5N1 has dramatically impacted wildlife in the wake of its spread across the
470 world. While it has infected an unprecedented diversity of species, we found that a
471 family-level phylogeny was sufficient to explain HPAI H5N1 notifications in wild birds,
472 potentially because ecological traits are often conserved across members of a family.
473 Using this same phylogeny to predict HPAI H5N1 notifications in Australian birds, where
474 the virus has yet to spread, we are able to predict that Sulidae, Gruidae, Laridae, and
475 Anatidae family members are likely to be most susceptible to disease from HPAI H5N1.
476 Similarly, we are able to predict susceptibility to disease in Australian endemic families,
477 such as the magpie goose (*Anseranas semipalmata*). Such predictions may provide
478 important support for those undertaking planning for potential HPAI H5N1 incursion
479 into Australia. Evaluating the accuracy of such predictions (and the method used to
480 generate them) will only be possible once (or 'if') HPAI H5N1 does indeed reach
481 Australian shores, and relies on continued and expanding monitoring efforts.

482

483

484

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499

500 **Author's contribution**

501 Study conceptualisation and design: SR, TR, MK; data collation: SR, TR, MK; data
502 analysis: SR with input from TR and MK; writing: SR with input from TR and MK.

503

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508

509 **Conflicts of interest/Competing interests**

510 The authors declare no conflicts of interest.

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