

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. HPAI transmission occurs through respiratory tracts and faecal-oral pathways,
69 which can take place directly through interaction with faecal matter or indirectly
70 through interaction with contaminated water, where the virus can persist for a long
71 time^{16,17}. Colony breeding, and specific colony traits such as distance between nests,
72 have also been implicated in HPAI H5N1 spread^{16,18}. Based on infection patterns in
73 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**

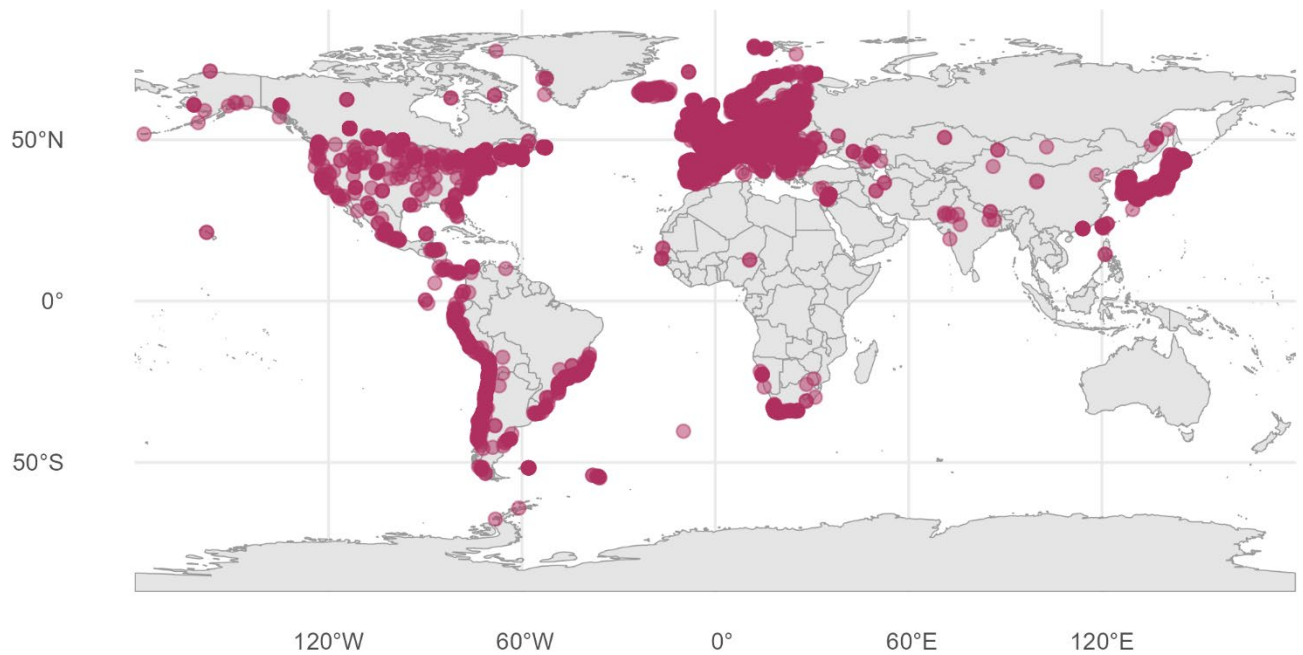
101 To model factors predicting HPAI H5N1 notifications in wild birds, we used the WOAAH
102 World Animal Health Information System (WAHIS) database of HPAI notifications.
103 WOAAH member countries are expected to submit information on a range of animal
104 diseases, and for each disease notification 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. We did not
121 filter the dataset to only include notifications where birds died. However, active
122 monitoring is less common in WAHIS reporting than passive monitoring (where testing is
123 done in response to visible disease), thus most notifications include diseased and dead
124 birds (only 3% of notifications reported 0 dead birds). While there is likely

125 underreporting that has important implications for the biases inherent in the dataset,
126 the WAHIS database still provides a minimum indication of HPAI notifications (and may
127 therefore underestimate, rather than overestimate, HPAI events). The database was
128 accessed on 16/02/2026 and filtered to only include notifications of outbreaks reported
129 in wild birds of known species since October 2021. In 95% of the cases, the subtype of
130 the HPAI notification was evaluated and established to be HPAI H5N1 (remaining being
131 H5 or unknown). We therefore assume that the majority (if not all) notifications used in
132 our study relate to HPAI H5N1, but will use the phrase 'HPAI' to capture the possibility of
133 other subtypes. It should be noted that WOAHA bears no responsibility for the integrity or
134 accuracy of the data, including but not limited to any deletion, manipulation, or
135 reformatting of data that may have occurred beyond its control.

136

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

146



147
 148 *Figure 1: Map of WAHIS HPAI notifications in wild birds. Each dot represents the*
 149 *reported location of a HPAI notification, and can represent one or several birds of a*
 150 *species. There is likely underreporting of HPAI notifications to WAHIS, but the data*
 151 *represents the minimum HPAI notifications. The data was accessed on 16/02/2026 and*
 152 *filtered to include notifications after October 2021.*

153
 154 Because of the links between disease transmission pathways and ecological traits like
 155 aquatic lifestyles, certain feeding ecologies, and tendency to congregate, we modelled
 156 how ecological traits might influence HPAI notifications in wild birds. Initial ecological
 157 categorisations of habitat and trophic niche were obtained from Avonet²⁵. In our
 158 analyses, we wanted to avoid categorisations that were too narrow (e.g. differentiating
 159 between frugivores and granivores) or perhaps arbitrarily differentiated between species
 160 with otherwise similar ecologies (e.g. denominating the Common Merganser as
 161 inhabiting “riverine” habitat, but other mergansers as “wetland”). Therefore, we
 162 modified some habitat and trophic niche categorisations based on information in Birds
 163 of the World²⁶, and broadened the groupings. We thus had three categories for habitat:
 164 Terrestrial, Freshwater, and Coastal/Marine, and four categories for diet: Predators
 165 (including vertebrate predators and scavengers), Piscivores, Plant-based diets
 166 (including aquatic and terrestrial plant material), and Omnivores (including any species
 167 that were both predators and plant-based feeders, and invertebrate predators as these

168 often also take plant matter). We used BirdLife’s list of congregating birds as our initial
169 starting points for whether species were known to congregate or not (Y/N), and
170 supplemented this with information from Birds of the World²⁶. For a full list of species
171 and our ultimate ecological categorisation for these species, see Supporting
172 Information Table S1.

173

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

183

184 *Statistical analyses*

185 All analyses were conducted in R version 4.4.0²⁷.

186

187 To model HPAI notifications in birds, we used a phylogenetic generalised linear mixed
188 model (GLMM) in the ‘brms’ package²⁸. In all the following models run in ‘brms’, we used
189 the flat default priors and ran models for 5000 iterations, with 600 iteration warmup and
190 a thinning interval of 5. Model convergence was assessed by visual inspection of chains
191 and by posteriors predictive checks of the models (using the ‘pp_check’ function). We
192 used a Poisson distribution of the number of HPAI notifications per species per region
193 (region defined using the ‘countrycode’ package as outlined above), controlling for
194 region as a fixed effect and running the model under a phylogenetic framework. For the
195 phylogeny, we used the family-level phylogeny from Kuhl et al²⁹. We used family, rather
196 than species-level phylogeny to avoid reporting biases for more common species, when
197 species in the same family are likely to share similar ecological traits and immune
198 system architecture. This was especially relevant for our next step, outlined below,
199 wherein we used the model of HPAI notifications to predict HPAI H5N1 disease

200 susceptibility in Australian species (we wanted to avoid drastically uneven outbreak
201 notification estimates for Australian species in the same family, but where some
202 species were closely related to a species with high HPAI notifications). We built upon
203 the phylogenetic GLMM to include the ecological traits of species: habitat (N = 3
204 categories), diet (N = 4 categories), and whether the species is known to congregate
205 (Y/N). The models did not include any offsets. We fitted three models that had one
206 ecological predictor (habitat, diet, or congregation) in addition to region as fixed effects,
207 and then an additional model that included all 3 ecological predictors and region. We
208 evaluated model fit of these against the null, region and phylogeny-only model using
209 leave-one-out (LOO) cross validation information criterion (IC), which is interpreted
210 similarly to AIC where low values are associated with better models.

211

212 The next step in our analyses was to predict which Australian species may be
213 susceptible to HPAI H5N1 once (or 'if') it arrives in Australia, based on patterns of HPAI
214 notifications elsewhere in the world. We used the HPAI notification data to predict
215 numbers of HPAI notifications in Australian birds, and use this as our metric of
216 predicted susceptibility to disease from HPAI H5N1. Most (~97%) of HPAI notifications
217 in the WAHIS database since October 2021 report deaths for species that have outbreak
218 notifications, which means that our predicted susceptibility is also linked to a species'
219 likelihood of experiencing sickness and death. To predict susceptibility to HPAI H5N1,
220 we first added the Australian bird families to our above phylogeny²⁹, thus resulting in a
221 phylogeny with the families in the WAHIS database and the Australian families. Using
222 the 'castor' package³⁰, we predicted HPAI notification likelihood onto the Australian
223 species. This was done using hidden state prediction via phylogenetic independent
224 contrasts, using the family-level phylogeny with both known and unknown HPAI
225 notifications (wherein known HPAI notifications were expressed as an average per
226 family). Through this, we retrieved predicted HPAI notifications for Australian bird
227 families, which we interpreted as their predicted susceptibility to disease from HPAI
228 H5N1.

229

230 Plots were made using 'ggtree'³¹ and 'ggplot2'³². Lastly, we extracted IUCN Redlist status
231 for Australian species using the 'rredlist' package³³, to ascertain the conservation status

232 of any species predicted to be highly susceptible to HPAI and thus highlight species that
233 may be at greater risk due to pre-existing vulnerabilities or due to other reasons than
234 vulnerability to HPAI H5N1.

235

236 *Animal ethics statement*

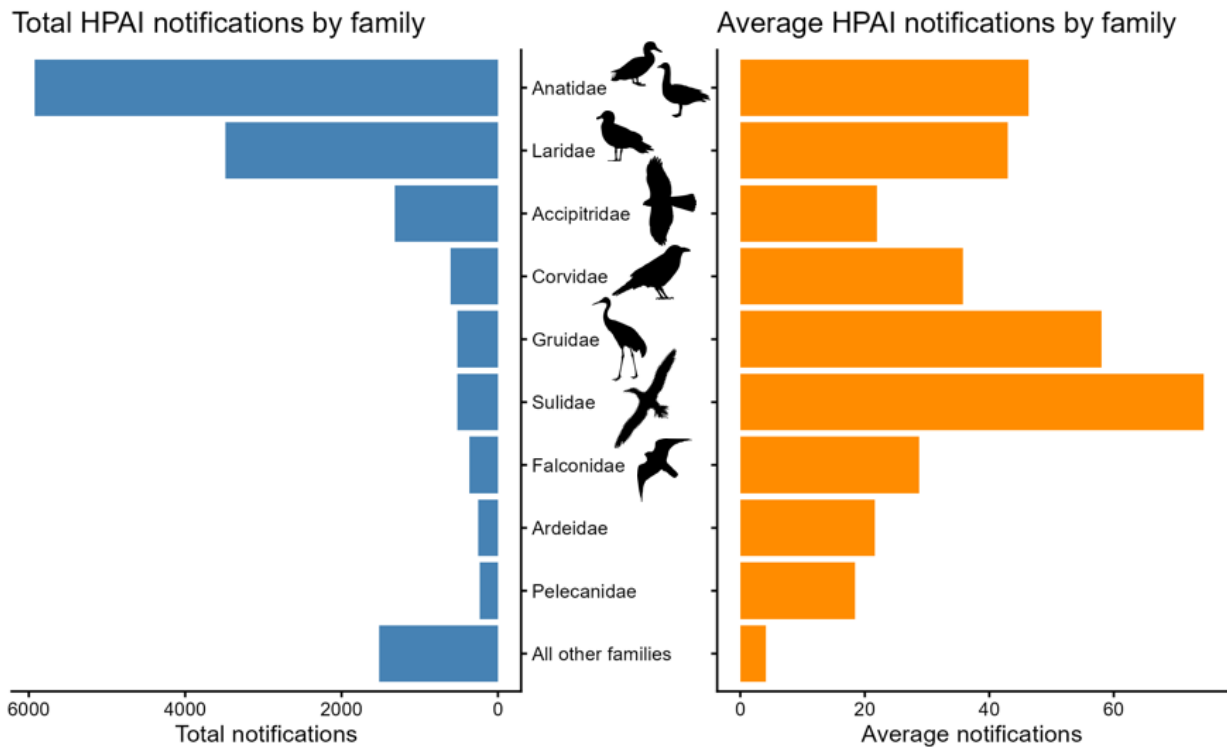
237 Our data is sourced from the WOAHS WAHIS database of global HPAI notifications in wild
238 birds. Therefore, our analysis is conducted on a pre-existing dataset with no new data
239 collected for the purpose of this study. We have no permit details to report.

240

241 **Results**

242 When analysing notifications of HPAI in wild birds since October 2021 using the
243 phylogeny-only model, we found a statistically significant phylogenetic signal (Pagel's λ :
244 0.56, 95% CI: 0.22 – 0.84). There were predominantly high numbers of total and average
245 HPAI H5N1 notifications amongst Anatidae (ducks, geese, and swans), Laridae (gulls,
246 terns, and noddies), Sulidae (gannets and boobies), and Gruidae (cranes). To a lesser
247 extent, crows and ravens (Corvidae), birds of prey (like Falconidae and Accipitridae),
248 and other seabirds (like Pelecanidae [pelicans] and Alcidae [auks]) also had higher
249 numbers of HPAI notifications (Figure 2).

250



251

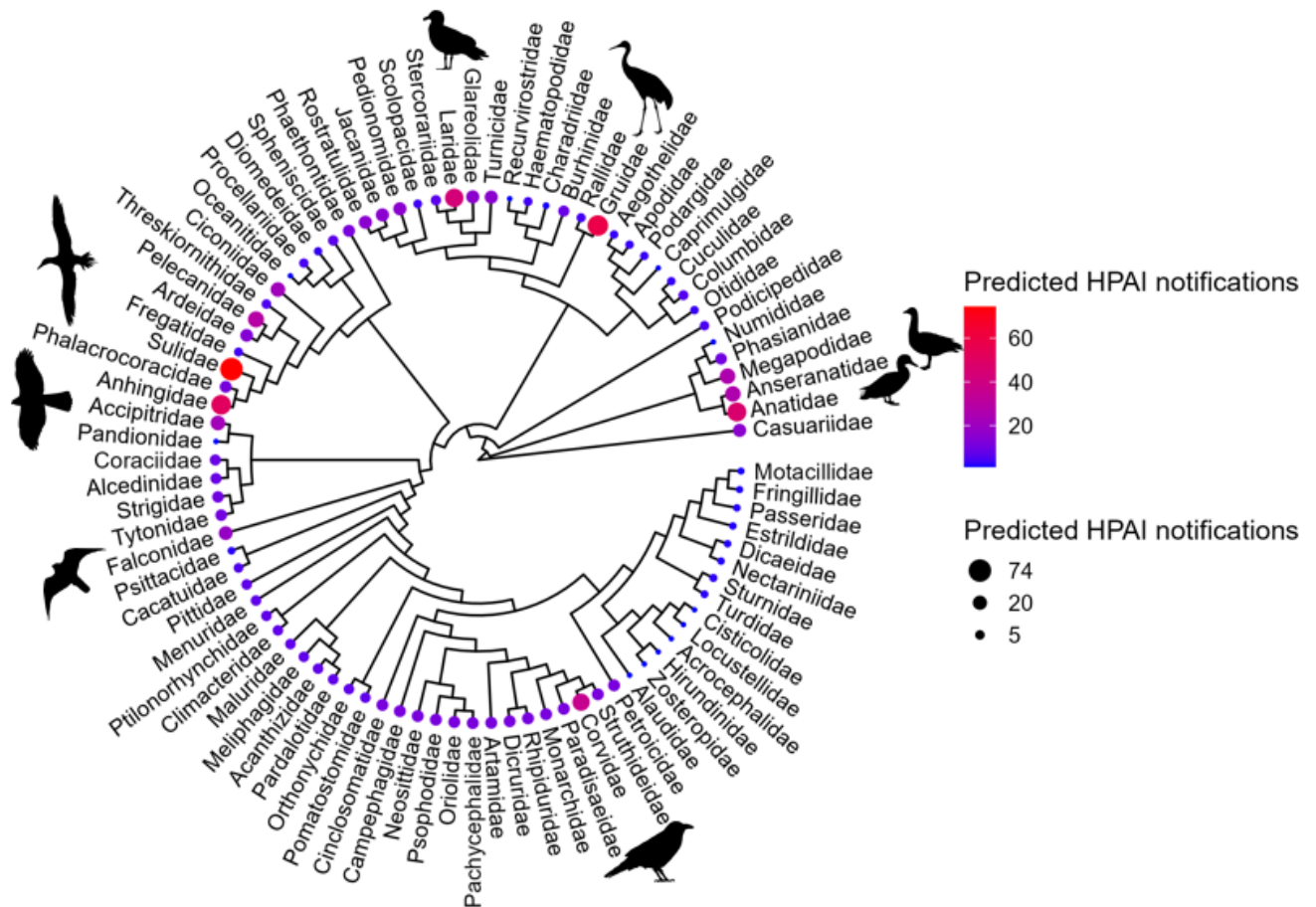
252 *Figure 2. HPAI notifications in wild birds 2021 – 2024. The left panel shows total HPAI*
 253 *notifications made to WOA/WAHIS per family, while the right panel shows average*
 254 *HPAI notifications per family. A few key families are highlighted by inclusion of bird icons*
 255 *from phylopic.org,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

257

258 Using leave-one-out (LOO) cross validation, we compared the fit of this null, region and
 259 phylogeny-only model and that of the ecological traits models. There were 4 models
 260 with ecological traits, where 3 models consisted of a single ecological trait (habitat,
 261 diet, or congregation) and a fourth model that included all three ecological traits (see
 262 Table S3 for summary statistics of all models). All four of these models included the
 263 phylogeny and region (see Methods). There was substantial overlap in the standard
 264 errors of the LOO ICs computed for the models, which means that the additional
 265 variables in our ecological models did not significantly improve model fit over the null,
 266 phylogeny-only model (Figure S1;³⁴). Therefore, we present results for the simpler,
 267 phylogeny-only model.

268

269 When predicting Australian species' HPAI H5N1 disease susceptibility (defined as their
270 predicted HPAI notifications), the predominant groupings of predicted notifications
271 were similar to that of the training model (Figure 3). Based on global HPAI notification
272 data since 2021 and the family-level phylogeny, the highest predicted HPAI H5N1
273 susceptibility was predicted for Australian Sulidae (gannets and boobies), followed by
274 Gruidae (cranes), Anhingidae (darters), Anatidae (ducks, geese, and swans), and
275 Laridae (gulls, terns, and noddies). Specifically, the model predicted 74 notifications
276 (i.e. HPAI detections in environmental, living, or dead bird samples) of HPAI in Sulidae
277 family members, followed by 58 in Gruidae, 51 in Anhingidae, 46 in Anatidae, and 43 in
278 Laridae family members. Other Australian bird families, like Pelecanidae (pelicans),
279 Corvidae (crows and ravens), and Accipitridae (falcons, hobbies, and kestrels), were
280 also predicted to be susceptible to disease. Furthermore, some families endemic to
281 Australia, such as Anseranatidae (containing the magpie goose – *Anseranas*
282 *semipalmata*), were predicted to be moderately susceptible to HPAI with a predicted 26
283 notifications. Predicted HPAI notifications for all Australian bird families in our list are
284 reported in Supporting Information Table S4.
285



286

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

292

293 Discussion

294 HPAI H5N1 2.3.4.4b has caused a panzootic of unprecedented scale⁵, but has not yet
 295 spread to Australia⁴. Here, we modelled HPAI notifications as a function of ecology and
 296 family-level phylogeny, finding that family-level phylogeny best explains number of HPAI
 297 notifications. The importance of host phylogeny in explaining avian influenza prevalence
 298 has been previously noted for low pathogenicity viruses¹⁵ and for HPAI H5N1 in different
 299 regions (e.g.^{10,12}), which we now expand for a global dataset of HPAI. Furthermore, we
 300 use phylogeny to predict HPAI notifications in Australian birds (including for Australian
 301 endemic birds), as a metric of susceptibility to HPAI H5N1 infection and resulting
 302 disease once (or ‘if’) it reaches the continent.

303

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

334

335 In our prediction of HPAI H5N1 susceptibility in Australian birds, we define susceptibility
336 as the predicted number of HPAI notification for a taxonomic family (where high
337 numbers of predicted HPAI notifications is interpreted as high susceptibility to disease;
338 with 74 HPAI notifications in Sulidae being the highest score). This modelling is based
339 on data of “outbreak notifications” from WOA H WAHIS since October 2021 (the onset of
340 the current panzootic), where most (~97%) species with notifications also have
341 reported deaths from HPAI. As such, assuming that notifications to WOA H WAHIS is
342 strongly correlated with disease and/or death, our predicted HPAI H5N1 susceptibility
343 largely reflects how easily different birds become infected and subsequently die of
344 HPAI. However, our susceptibility predictions largely ignore the role of different birds in
345 maintaining and spreading HPAI H5N1, since different species might survive (and some
346 may indeed not die from infection) and carry the virus for different lengths of time. For
347 example, bald eagles (*Haliaeetus leucocephalus*) had higher HPAI H5N1
348 seroprevalence (indicating higher survival rate) than other birds of prey³⁶, and anti-H5
349 antibodies in seabird eggs were higher in common eiders compared to other seabird
350 species (such as gannets, which suffered HPAI-related mass mortality events³⁷). In the
351 current HPAI panzootic, recent research has shown that the host dynamics of H5N1
352 differs between virus genotypes³⁸, showcasing the wide range of birds capable of
353 contributing to the spread of HPAI and that the reservoir community can change rapidly.
354 Furthermore, migratory and partially-migratory birds are likely to have a significant role
355 in sustained transmission of HPAI H5N1 to other birds, as has been found in other
356 regions^{10,12,13}. This suggests that species can play different roles in maintenance and
357 spread of HPAI H5N1 after exposure to the virus, which is important to consider when
358 predicting HPAI H5N1 susceptibility. How HPAI H5N1 spread may look like within the
359 Australian continent is further complicated by the often nomadic movements of
360 Australian waterfowl as they move in response to ephemeral wetlands³⁹.

361

362 In this study, we used a family-level phylogeny to avoid biases associated with
363 particular outlier species. Our approach may still carry some inherent biases, for
364 example if a family is very speciose, very abundant, or contains very commonly
365 sampled species. However, when comparing the mean number of HPAI notifications per
366 family to the number of species in that family, the correlation was low ($R^2 = -0.06$),

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

383

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

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

412

413 In our analysis, adding ecological traits like diet and habitat did not significantly improve
414 the predictability of HPAI notifications above our null, phylogeny-only model. However,
415 this does not mean those traits are not still important to consider when assessing virus
416 incursion into new ranges, like Oceania. For example, traits like colony-breeding can
417 amplify the risk to a species if the virus is able to spread rapidly through a large number
418 of birds¹⁸, but this also means there is a temporally-stratified risk to consider as colony-
419 breeding is only a relevant trait at certain times of the year²¹. Australia hosts big
420 breeding colonies of gannets, shearwaters, and other notable seabirds, which might
421 expose them to the same colony-wide mass mortalities noted elsewhere^{6,40,41}.

422 Therefore, even if dense flocking behaviour was not a major predictor of HPAI
423 notifications in our analysis, it is a trait worth bearing in mind when considering
424 conservation impacts of potential HPAI H5N1 arrival in Australia. The same holds for
425 diet and habitat traits, which were kept deliberately broad in our analysis. For example,
426 the ecologies of geese and passerines were in this analysis occasionally grouped as
427 ‘terrestrial omnivores’, but their distinct feeding ecologies and habitat movements can
428 mean they will play very different roles in transmission of an HPAI H5N1 incursion into
429 Australia. Similarly, analysis of the geospatial distribution of HPAI H5N1 has revealed

430 that landscape characteristics can also influence risk²¹, which will be crucial to
431 consider when predicting potential incursion and spread in the Australian continent.

432

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

449

450 Among other factors that can be important to consider when predicting HPAI H5N1
451 susceptibility in Australian birds is conservation status, where rampant disease spread
452 may have a larger impact on more vulnerable populations²². In Australian species with
453 highest predicted HPAI notifications (within the top 50 predicted HPAI H5N1
454 notifications), all but three are listed as Least Concern on the IUCN Red List status.
455 Only the fairy tern (*Sternula nereis*), sooty tern (*Onychoprion fuscatus*), and the sarus
456 crane (*Grus antigone*) are listed as Vulnerable. However, expanding this to species
457 predicted to be moderately susceptible to HPAI (top 80 predicted disease notifications),
458 there is one species listed as Endangered (red goshawk – *Erythrotriorchis radiatus*), two
459 additional birds listed as Vulnerable (grey falcon – *Falco hypoleucos* and malleefowl –
460 *Leipoa ocellata*), and one Near Threatened (letter-winged kite – *Elanus scriptus*). It is
461 notable that most of these are birds of prey, suggesting that while HPAI H5N1 is

462 primarily predicted to impact waterbirds, the notable conservation impacts of potential
463 HPAI H5N1 incursion into Australia may focus on predators. The impact to predators
464 might be similar to effects seen elsewhere²¹, such as declining peregrine falcon (*Falco*
465 *peregrinus*) populations in the Netherlands, where over 80% of tested dead birds were
466 infected with HPAI H5N1⁴⁴. Conservation vulnerability of predators to HPAI H5N1 also
467 underscores the potential conservation concerns to mammalian predators¹⁹, as has
468 been noted in South American pinnipeds⁴⁵.

469

470 **Conclusion**

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

484

485

486

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501

502 **Author's contribution**

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

505

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510

511 **Conflicts of interest/Competing interests**

512 The authors declare no conflicts of interest.

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