¹ Wild birds' plight and role in the current bird flu panzootic

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13 Abstract

14 The current avian influenza panzootic is unprecedented and catastrophic for birds. With a focus on

the implications of this panzootic for poultry, there is limited attention on wild birds. We highlight

16 shortcomings and geographic biases in reporting leading to a severe underappreciation of wildlife

17 mortality. We estimate the scale of mortality amongst wild birds is in the millions rather than tens-

18 of-thousands reported, through comparison of notification data to accounts literature. The

19 outbreaks amongst wild birds are causing population and species level concerns which may drive

20 extinctions and jeopardise decades of conservation efforts.

21

22 **Main**

- 23 Since its emergence in Hong Kong in 1996, high pathogenicity avian influenza (HPAI) viruses of
- the A/Goose/Guangdong/1/96 lineage have steadily evolved. Since 2014, they have become
- 25 increasingly widespread and progressively more destructive across the poultry industry and wild
- ²⁶ birds, globally ¹. This has culminated in the emergence of the descendant lineage 2.3.4.4b, which
- since October 2021 has spread to all continents except the Antarctic and Australia, resulting in a

28 panzootic of unprecedented magnitude (Fig 1A-B). According to the World Animal Health

29 Information System (WAHIS) from the World Organisation for Animal Health² this panzootic has

30 resulted in the death and destruction of more than half a billion poultry. Wildlife, notably birds, have

- also taken a profound hit. Based on available data, 2.3.4.4b has infected and killed a substantially
- 32 greater diversity of wild birds (i.e., 320 species belonging to 21 orders) compared to previous
- 33 lineages (Fig. 1 C). Moreover, while prior to October 2021 most HPAI cases in wild birds occurred
- 34 near intensive poultry production, recent outbreaks in wild birds now also include remote areas
- thousands of kilometres away from poultry production.

36 While the reported scale of poultry mortality due to HPAI is likely aligned to reality, the 68,013 wild bird casualties and 289 mammalian cases reported in WAHIS² are not. In part, this is due to 37 inconsistent reporting, either including all observed carcasses, only those tested and confirmed, or 38 no information on numbers at all. For instance, the outbreaks comprising 24,463 Cape Cormorants 39 40 in South Africa (*Phalacrocorax capensis*) and 9,029 Common Cranes (*Grus grus*) in Israel appear to be reported in full. Apparently more widespread is the reporting of fractions of cases compared 41 to what is stated in the media³, or from publications. For example, in the WAHIS database there 42 are 41 HPAI notifications for Sandwich Terns (Thalasseus sandvicensis) involving 68 individuals 43 globally, since 1 October 2021. This contrasts sharply with 9,600 dead Sandwich Terns ⁴ reported 44 in the Netherlands alone, of which only 19 are included in WAHIS. In the media a further 7000 45 Sandwich Tern carcasses were reported in France, the UK and Germany, which together 46 comprises a two to three order of magnitude difference between official notifications in WAHIS and 47 48 reports from the ground. Similarly, 73 dead Great Skuas (Stercorarius skua) have thus far been 49 reported to WAHIS, while 1400 dead individuals were found on the small Scottish island of Foula alone ⁵. Additionally, 593 Dalmatian Pelicans (*Pelecanus crispus*) were reported to WAHIS from 50 Greece, whereas 2,286, or approximately 40% of the SE European population, have been reported 51 in the literature ⁶. Finally, in some cases, outbreaks are reported only in the literature and not in 52 WAHIS, such as the 2140 Great White Pelicans (Pelecanus onocrotalus) found dead in Mauritania 53

54 in early 2021 7 .

55 Humans report outbreaks, and the correlation between the distribution of wildlife cases reported

and human population density strongly suggests the potential for vast underreporting in areas

57 where few people live (Figure 1D). Combined with the high underreporting of outbreaks in areas

where human population densities are still high (e.g., the case of the Dutch Sandwich Terns),

59 suggests only a fraction of outbreaks in wildlife have been detected and appropriately reported.

60 Therewith, the number of wild birds impacted is conceivably in the millions rather than the tens of

thousands that have been reported and collated in WAHIS 2 .

62 If HPAI continues to frequently spill over from poultry to wildlife, wild bird populations will continue to be affected. The global poultry population currently comprises 70% of the world's avian biomass 63 ⁸ and plavs a central role in the perpetuation of HPAIs ⁹. Until recently, HPAI outbreaks in wild birds 64 65 could be directly linked to spill-over from poultry, and wild birds did not play a central role in virus perpetuation. For example, HPAI repeatedly disappeared after annual incursions in Europe 66 between 2005 and 2008, and North America and Europe in 2014¹⁰. However, now HPAI 2.3.4.4b 67 has adapted from poultry to wild birds ¹¹, explaining the explosion of sustained outbreaks in wild 68 69 birds since 2020, including outbreaks in remote areas with no poultry.

70 Despite not historically being the main reservoir for HPAI, wild birds play a key role in long distance

viral spread. This is facilitated by individuals that show only mild or no clinical signs after virus

infection. Tracking studies in China have shown that Mallards (Anas platyrhynchos) are capable of

- flying hundreds of kilometres while infected with HPAI 2.3.4.4b¹². These birds' tolerance for HPAI
- infection is likely explained by their infection history. Waterfowl typically have high rates of low
- pathogenic avian influenza prevalence, including exposure to low pathogenic H5 strains ¹³. These
- ⁷⁶ infections may provide homo- or heterosubtypic immunity against subsequent HPAI infections ¹⁴,
- preventing infection or dampened disease severity ¹⁴, serving as one explanation how such virulent
- virus can spread so readily on-board wild birds.

Given the enormous burden of this virus on poultry and wild birds alike, there are increased calls

- 80 for poultry vaccination. While proven successful in significantly reducing mortality ¹⁵, vaccination of
- 81 poultry may also drive virus evolution ¹⁶, leading to continued virus circulation in vaccinated flocks
- 82 with limited disease signs¹⁷, contributing to endemicity and spill over risk to wild birds. However,
- 83 while not a silver bullet, when done in combination with monitoring to guarantee disease freedom,
- 84 a "vaccination plus strategy" should ideally decrease not only disease burden on poultry, but also
- on wild birds and mammals by limiting spill-over from poultry production.

In summary, despite lacking an accurate estimate of the true impact on wildlife, we are witnessing

- 87 a panzootic of an unprecedented and enormous scale. This panzootic did not emerge from
- nowhere, but rather is the result of 20 years of viral evolution in the ever-expanding global poultry
- population. Given the key role of poultry production in food chains, and the effect on livelihoods, it
- 90 is logical for countries to prioritize their response towards poultry. However, that wild bird outbreaks
- are widely neglected, to the degree that we do not even know the order of magnitude of deaths,
- nor the population and ecosystem consequences, is highly concerning. As a result, the true impact
- of this panzootic on wild birds may not be recognised for years to come, and some species may
- 94 never recover.



95

Figure 1. (A) HPAI H5 outbreaks in wild birds and poultry in the two waves prior to the current panzootic. (B) HPAI H5 outbreaks during the current panzootic. (C) Number of reports of HPAI in wild birds, per half year, across avian orders (stacked bars, left axis), and the number of wild bird species involved (brown line, right axis). (D) Histogram of the number of reports received on wild birds infected with HPAI since 2005 as a function of human population density (on a 10-log scale) compared to a histogram depicting the global distribution of human density across the land masses (wing a one by one depicted end lengitude and lengitud

102 (using a one-by-one degrees of latitude and longitude gridded map) (orange).

103

104 **References**

105	1	Wille, M. & Barr, I. G.	Science 376,	459-460, doi:10.1126/science	e.abo1232 (2	2022)
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- 106 2 WOAH. <u>https://wahis.woah.org/</u> (2023).
- 107 3 Clough, C. Veterinary Record **191**, 221-222, doi:10.1002/vetr.2214 (2022).
- 108 4 Rijks, J. M. *et al. Emerg Infect Dis* **28**, 2538-2542, doi:10.3201/eid2812.221292 (2022).
- 109 5 Camphuysen, K. & Gear, S. *NIOZ report*, doi:10.25850/nioz/7b.b.gd (2022).
- 110 6 Alexandrou, O., Malakou, M. & Catsadorakis, G. *Oryx* **56**, 813-813,
- 111 doi:10.1017/S0030605322001041 (2022).
- 112
 7
 Beyit, A. D. et al. Veterinary Research Communications, doi:10.1007/s11259-023-10100-6

 113
 (2023).
- Bar-On, Y. M., Phillips, R. & Milo, R. *Proceedings of the National Academy of Sciences* **115 115**, 6506-6511, doi:doi:10.1073/pnas.1711842115 (2018).
- Gilbert, M., Xiao, X. & Robinson, T. P. Arch Public Health 75, 48, doi:10.1186/s13690-017 0218-4 (2017).
- 118 10 Krauss, S. *et al. Proc Natl Acad Sci U S A* **113**, 9033-9038, doi:10.1073/pnas.1608853113 119 (2016).
- 120 11 James, J. *et al. bioRxiv*, 2023.2002.2007.527270, doi:10.1101/2023.02.07.527270 (2023).
- 121 12 Lv, X. et al. Emerging Infectious Disease journal **28**, 1039, doi:10.3201/eid2805.211580 (2022).

- 123 13 Latorre-Margalef, N. et al. Proceedings of the Royal Society B: Biological Sciences 281,
- 124 20140098, doi:10.1098/rspb.2014.0098 (2014).
- 125 14 Tarasiuk, K. *et al. Veterinary Research* **53**, 108, doi:10.1186/s13567-022-01125-x (2022).
- 126 15 Sims, L. D. Avian diseases **51**, 174-181 (2007).
- 127 16 Gandon, S., Mackinnon, M., Nee, S. & Read, A. *Proceedings of the Royal Society of London. Series B: Biological Sciences* 270, 1129-1136, doi:10.1098/rspb.2003.2370
 (2003).
- 130 17 Salaheldin, A. H. *et al. Viruses* **14**, doi:10.3390/v14071431 (2022).
- 131

Online Methods

133 Data were downloaded from the World Organization for Animal Health (WOAH) World

- 134 Animal Health Information System (WAHIS) database on 05/03/2023. Data extracted by
- 135 Marcel Klaassen, Deakin University. Reproduced with permission. WOAH bears no
- responsibility for the integrity or accuracy of the data contained herein, but not limited to,
- any deletion, manipulation, or reformatting of data that may have occurred beyond itscontrol.
- 139 Special thanks to Paolo Tizzani from WOAH for kindly providing additional information on
- this data source.
- 141 "Al indeterminatum fau" are assumed to be birds.
- 142

```
knitr::opts chunk$set(echo = TRUE)
143
144
      library(raster)
145
      library(rgdal)
146
      library(sp) # used to create a SpatialPoint object
147
      library(readxl)
148
      library(ggplot2)
149
      library(scales) #change labels along axis from scientific notation to
      "ordinary"
150
151
152
      #get WOAH data
      I <- read excel("data/infur 20230303.xlsx")</pre>
153
      B <- read_excel("data/BirdListI.xlsx")</pre>
154
155
156
      I$Species[which(I$Species=="Hieraaëtus fasciatus")] <- "Hieraaetus fasciatus"</pre>
157
158
      #retain only avian influenza
      I <- I[which(I$disease eng %in% c("Influenza A viruses of high pathogenicity
159
      (Inf. with) (non-poultry including wild birds) (2017-)", "High pathogenicity
160
      avian influenza viruses (poultry) (Inf. with)")),]
161
162
163
      #if observation date is missing take reporting date instead
164
      I$Date <- I$Outbreak start date
165
      I$Date[which(is.na(I$Date))] <- I$Reporting_date[which(is.na(I$Date))]</pre>
166
      #only cases with a date
167
168
      I <- I[which(!is.na(I$Date)),]</pre>
169
170
      #make sure lats and longs are numeric
171
      I$Latitude <- as.numeric(I$Latitude)</pre>
```

```
172
      I$Longitude <- as.numeric(I$Longitude)</pre>
173
174
      SpeciesListI <- unique(I$Species)</pre>
      SpeciesListB <- unique(B$Species)</pre>
175
176
      I <- merge(I,B, by="Species", all.x=TRUE)</pre>
177
178
      #retain only wild animals
179
      I <- I[which(I$is wild==TRUE),]</pre>
180
181
      #retain Aves
```

182 I <- I[which(I\$Class=="Aves"),]</pre>

Finding wild birds affected by HPAI may be a function of the number of people living there. Large colonies of (sea) birds can often be found in remote areas. The question thus arises what we actually do pick up of what is happening to wild birds globally?

- 186 World population density data was downloaded from
- 187 http://sedac.ciesin.columbia.edu/data/collection/gpw-v4/documentation and plotted un-
- transformed and on a log scale. From these maps and for all HPAI incidence data we
- 189 extracted the corresponding human population density of which next a histogram was
- 190 plotted. This histogram suggests that the majority of reports come from areas where human
- 191 population densities are at an intermediate level. This pattern may be due to the fact that
- where many people live few birds occur in high densities and where few people live there
- 193 may be affected birds, but they are less likely to be noted and reported.
- 194 Further support for this pattern emerges when studying the outbreak maps and focusing
- 195 on the northern hemisphere shorelines of temperate and Arctic regions. These are
- 196 characterised by large seabird colonies in which large outbreaks have been recorded over
- the past year, but only in those areas where population densities are relatively high
- 198 (Svalbard, Iceland, Europe, west coast of Canada and south coast Alaska) but not
- 199 elsewhere (north coast North America, Greenland, Russia). It should also be noted that
- 200 due to the COVID-19 pandemic research activities in these regions have also been limited.
- There is thus a very high chance that the size of the AIV panzootic is considerably underreported.
- 203 #get the population density data 204 p <- brick("data/gpw_v4_population_density_rev11_1_deg.nc") 205 206 # the 2020 data are at layer 5 207 spplot(p[[5]], main = "GPWv4 Population Density, v4.11_2020")





208

209 spplot(log(p[[5]]), main = "GPWv4 ln(Population Density), v4.11_2020")



GPWv4 In(Population Density), v4.11_2020

210

```
211
      I$Pop <- NA
212
      #get the population density for each outbreak location
213
      for (i in 1:nrow(I)) {
214
        I$Pop[i] <- raster::extract(x=p[[5]],y=I[i,c(which( colnames(I)=="Longitude"</pre>
215
      ),which( colnames(I)=="Latitude" ))])
216
        }
217
      #get reference population density for each outbreak location
218
      #create data frame with 0 rows and 3 columns
219
      df <- data.frame(Longitude=numeric(0),Latitude=numeric(0))</pre>
220
221
      i <- 0
222
      for (Longitude in -180:180) {
223
        for (Latitude in -90:90) {
224
          i <- i+1
225
          df[i,] <- c(Longitude,Latitude)</pre>
226
        }}
227
      df$Pop <- NA
228
      #get the population density for each outbreak location
229
      for (i in 1:nrow(df)) {
```

```
230
        df$Pop[i] <- raster::extract(x=p[[5]],y=df[i,c(which(</pre>
231
      colnames(df)=="Longitude" ),which( colnames(df)=="Latitude" ))])
232
        }
      #remove where there is no land
233
234
      df <- df[which(!is.na(df$Pop)),]</pre>
235
236
237
      #make hist of outbreak frequency as function of pop density
238
239
      Fig1D <- ggplot() +</pre>
240
        geom_histogram(data=df,aes(x=Pop),colour="darkorange", fill="darkorange",
241
      alpha=0.5) +
        geom histogram(data=I,aes(x=Pop),colour="darkred", fill="darkred", alpha=0.5)
242
243
      +
244
        scale x log10(labels=comma format(digits=3)) +
245
        labs(
             x=" human population density (per square km)") + theme classic()
246
247
248
      plot(Fig1D)
```



249 250

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