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# The Three Ts of Virulence Evolution During Zoonotic Emergence

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## Abstract

There is increasing interest in the role that evolution may play in current and future pandemics, but there is often also considerable confusion about the actual evolutionary predictions. This may be, in part, due to a historical separation of evolutionary and medical fields, but there is a large, somewhat nuanced body of evidence-supported theory on the evolution of infectious disease. In this review, we synthesize this evolutionary theory in order to provide framework for clearer understanding of the key principles. Specifically, we discuss the selection acting on zoonotic pathogens' transmission rates and virulence at spillover and during emergence. We explain how the direction and strength of selection during epidemics of emerging zoonotic disease can be understood by a three Ts framework: trade-offs, transmission, and time scales. Virulence and transmission rate may trade-off, but transmission rate is likely to be favored by selection early in emergence, particularly if maladapted zoonotic pathogens have 'no-cost' transmission rate improving mutations available to them. Additionally, the optimal virulence and transmission rates can shift with the time scale of the epidemic. Predicting pathogen evolution

42 therefore depends on understanding both the trade-offs of transmission-improving mutations and  
43 the time scales of selection. (194/200)

44

45 **Keywords:** Trade-offs, Virulence, Transmission, Emerging Zoonotic Disease, Evolution

46

## 47 **1. Introduction**

48 Throughout the current global pandemic of Sars-CoV-2, we have seen a growing public  
49 fascination with the role of pathogen evolution during disease emergence. In May 2020, reports  
50 of a mutational variant (D614G) increasing in frequency sparked concern about virus evolution [1]  
51 and more potentially adaptive variants have since been reported [2]. These experiences with  
52 SARS-CoV-2 and with previous epidemics of other zoonotic diseases have clearly demonstrated  
53 the potential for pathogens to evolve during disease emergence [3]. Despite this importance,  
54 public conversations around pathogen evolution are often fraught with misunderstandings. To  
55 some extent, this is likely reflective of the historical separation of evolutionary and medical  
56 disciplines [4]. Beyond that, however, scientific communication around pathogen evolution is  
57 particularly tricky because the science to be communicated provides no clear answers to be  
58 packaged into simple explanations.

59 Experts studying infectious disease evolution understand that pathogens have the  
60 potential to rapidly adapt due to high population sizes, short generation times, and relatively high  
61 mutation rates [5] and recognize that human populations impose novel, although often  
62 understood, selection pressures [6]. At the same time, however, many experts are sometimes  
63 quick to express skepticism when public conversation is dominated by concern over pathogen  
64 evolution. This is partially because pathogen evolution is just one factor of many that collectively  
65 influence epidemic progression, so communication around its importance sits on a teeter totter of  
66 balancing a concern and attentiveness against a blinded focus on potential evolution over other  
67 factors shaping the epidemic [7,8].

68 Additionally, many experts studying infectious disease evolution are often quick to  
69 emphasize that we cannot predict how a specific pathogen will evolve [9]. This, however, does  
70 not mean that we have absolutely no idea of how pathogens generally may evolve. We expect  
71 that pathogens will evolve in response to selection in human populations, but the speed at which  
72 they do depends critically on the availability of adaptive variation and the relative strength of  
73 selection compared to stochasticity, both of which relate to the number of infected individuals [10].

74 Theory predicts that pathogens may evolve towards optimal virulence and transmission rates due  
75 to underlying constraints, but these predictions depend on nuances of pathogen biology, epidemic  
76 stage, and host population structure [11,12]. It can, understandably, be frustrating when asking  
77 how a pathogen will evolve to hear predictions that sound like contradictions and non-answers,  
78 but this reflects the complicated realities of pathogen evolution. However, this real uncertainty  
79 also seems to have created an environment where hope for simple answers means that  
80 misinformation can spread.

81 On top of the inherent challenges of communicating complex scientific concepts,  
82 researchers studying pathogen evolution must also play ‘whack-a-mole’ against a variety of  
83 misconceptions that are wrong in different ways. Public concern sometimes skews towards  
84 pathogens evolving to be hyper-virulent, hyper-transmissible superbugs [13]. Alternatively,  
85 historical theories of evolution towards avirulence still pervade the public consciousness and  
86 sometimes lead to the prediction that pathogens universally evolve to become less dangerous  
87 [14]. In both directions, these misconceptions can lead to inappropriate public health policies.  
88 However, the disjointed nature of combatting misconceptions as they arise has led to much of the  
89 conversation on pathogen evolution in emerging zoonotic diseases being scattered across the  
90 scientific literature and media. This can be compounded by the fact that researchers studying  
91 pathogen evolution come from a variety of sub-disciplines and their work is often not well  
92 integrated [15].

93 As pathogen evolution continues to be an important conversation in the current pandemic  
94 of SARS-CoV-2 and is likely to again be important during future epidemics of emerging zoonotic  
95 disease, this review aims to collect insights from the wealth of research on pathogen evolution to  
96 provide a centralizing, conceptual understanding of the factors shaping the evolution of  
97 transmission rate and virulence in epidemics of novel zoonotic disease. While we cannot  
98 comprehensively discuss this vast literature, our aim is to provide a framework so that readers  
99 understand the general principles of pathogen virulence and transmission evolution and can also  
100 see how variations in the assumptions of these models based upon nuances of biology and  
101 population structure can lead to deviations in their predictions. Because strong reviews of  
102 virulence evolution exist elsewhere in the literature [4,12], our review focuses specifically on  
103 virulence evolution in epidemics of novel zoonotic disease to focus on how general theory for  
104 virulence evolution is altered by the specific characteristics of emerging zoonotic diseases and  
105 shifting selection pressures during epidemics. Extending beyond the scope of any single  
106 theoretical paper on this topic, we will discuss: (1) how do trade-offs between pathogen traits

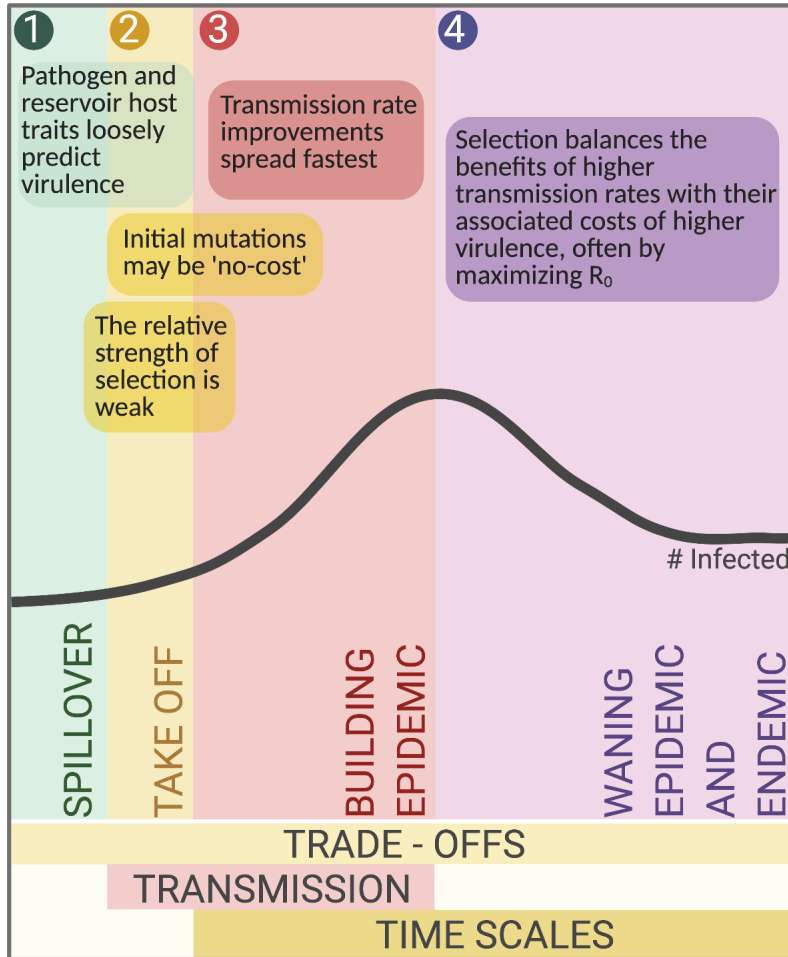
107 constrain pathogen evolution?; (2) what predicts pathogen virulence at the spillover barrier?; (3)  
108 why it is hard to predict how novel zoonotic pathogens will evolve?; and (4) how do optimal  
109 strategies in populations with different epidemiological characteristics change over time during an  
110 epidemic? Through this, we describe predictions for pathogen evolution during epidemics of  
111 emerging zoonotic disease and how they can change depending on pathogen biology and host  
112 population structure.

113

## 114 **2. The Three Ts Framework: Trade-offs, Transmission, and Time Scales**

115 The adaptive evolution of any trait depends on the presence of variation and the ability of  
116 selection to act on that variation. It is clear that pathogens, particularly RNA viruses, can quickly  
117 generate and maintain large amounts of variation [16]. At the start of an epidemic, selection on  
118 these variants is weak compared to stochastic and demographic pressures, but gains strength as  
119 the number of infections increase [10]. Selection on virulence during epidemics of emerging  
120 zoonotic disease can be understood by considering the ‘three Ts’: trade-offs, transmission, and  
121 time scales [7,17–19]. See Figure 1 for graphical summary.

122 In terms of **trade-offs**, theory has often assumed, and empirical data has increasingly  
123 shown us, that many pathogen traits, like transmission rate and virulence, **trade-off** with each  
124 other [12,17,20,21] (See Table 1). The **trade-off** theory is important because it explains how  
125 different intermediate virulence, transmission, and recovery rates can be optimal for a pathogen  
126 due to constraints between these key traits [12,17,21]. In terms of **transmission**, emerging  
127 zoonotic pathogens typically do not have histories of selection in human populations and thus are  
128 likely to be maladapted for human-to-human transmission [22]. This maladaptation potentially  
129 means that emerging zoonotic pathogens may initially have ‘no-cost’ mutations available that  
130 improve transmission rate without impacting traits like virulence [18]. In these cases, emerging  
131 diseases can be selected to increase their **transmission rates** with no, or potentially  
132 counterintuitive, impacts on virulence [18]. Finally, **time scale** matters since, even with trade-offs  
133 between virulence and transmission rate, transmission rate improvements continue to be the most  
134 important selection pressure at the start of an epidemic because the relative strength of selection  
135 on transmission rate and virulence shifts as the density of susceptible hosts changes during an  
136 epidemic [19,23]. This effect further alters a number of theoretical predictions that are classically  
137 evaluated at equilibrium for how different host, pathogen, and epidemiological factors shape  
138 selection on pathogen traits. Therefore, a pathogen’s optimum strategy changes over **time** during  
139 an epidemic under a wide array of conditions. We will discuss each of these in detail below.



140

Figure 1: The Three Ts of Virulence Evolution During Zoonotic Emergence. Trade-offs between virulence and transmission rate determine pathogen fitness at every point during an epidemic, regulating pathogen fitness at the spillover barrier and shaping selection as the epidemic progresses. Early in the epidemic, however, individual transmission rate improving mutations may be 'costless' and not have trade-offs. Improvements in transmission rate are the most important selection pressure during epidemic take-off and building phases, though selection is weak at take-off. Finally, the time scale of the epidemic shifts the pathogen's optimal virulence and transmission rate strategies as the density of susceptible hosts changes. Created with Biorender.com

141 **3. How do trade-offs between pathogen traits constrain pathogen evolution?**

142 Evolutionary biologists have long been interested in why pathogens harm their hosts, or  
 143 cause virulence (Figure 2) [24]. Based on the assumption that host damage was detrimental to  
 144 parasite fitness, early ideas predicted that all parasites should evolve towards avirulence [4,14].  
 145 This was considered the 'conventional wisdom' until the 1980s, when foundational papers began  
 146 to appreciate that virulence might be linked to other parasite traits like transmission or recovery  
 147 rates and therefore could have an evolutionary optimum [17]. Trade-offs between these traits  
 148 would mean that low virulence would come at a cost of low transmission rate or fast recovery and

149 that avirulence would therefore hinder parasite fitness. This virulence and transmission trade-off  
150 is now fundamental to our theories on pathogen evolution.  
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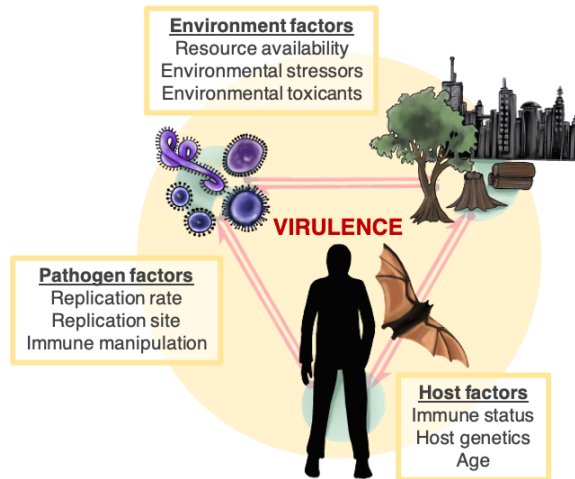


Figure 2: Disease Triangle of Virulence

152 Theory on the virulence and transmission trade-off typically suggests that virulence and  
153 transmission rate are both functions of the within-host exploitation or replication rate [4,12].  
154 Because faster replicating pathogens generate larger population sizes, they increase their  
155 transmission rate while causing more host damage [12,21]. Damage increases host mortality,  
156 thereby decreasing the host's infectious period and providing a shorter window for the infected  
157 host to contact susceptible hosts [17]. In short, faster within-host replication increases the  
158 likelihood of infection upon contact while decreasing the overall duration of infection [17,21].  
159 Under the trade-off hypothesis, parasites are therefore selected for exploitation rates that balance  
160 virulence and transmission rate [12,17,21].

161 Transmission rate and virulence do not necessarily need to trade off through the within-  
162 host exploitation rate for selection to balance the two traits. A virulence-recovery trade-off can  
163 occur if low replication rates make pathogens easier to clear such that lower virulence trades off  
164 with faster recovery rates [17]. Alternatively, a transmission-recovery trade-off can occur if the  
165 immune response is activated in a density dependent manner so that high replication rates have  
166 high transmission rates, but fast recovery [25]. A sickness behavior-transmission trade-off may  
167 result if faster replication rates make the host feel sick and isolate themselves so that high  
168 replication leads to a higher probability of infection upon contact, but fewer contacts [26]. Finally,  
169 the virulence and transmission trade-off does not necessarily depend on changes to the within-  
170 host replication rate if symptoms themselves are needed for transmission [27].

171 In simple host-parasite models, pathogens are selected to maximize the epidemiological  
 172  $R_0$  (i.e. the number of secondary infections that a parasite produces during its infectious period in  
 173 an entirely susceptible population) [17] (but see [28,29]). The virulence-transmission trade-off  
 174 predicts that these two traits are positively correlated, but the shape of this relationship is critical  
 175 to the predictions of evolutionary theory [17,21]. When the trade-off is linear, pathogens evolve  
 176 maximum virulence; but when the trade-off is saturating (such that virulence is acceleratingly  
 177 costly in terms of transmission rate), pathogens will evolve towards an intermediate virulence  
 178 [4,17]. Given the centrality of the trade-off hypothesis to our understanding of virulence, it is  
 179 noticeable that there are a number of empirical studies that have found support for the core idea  
 180 (See Table 1, Rows 1-2) [20].  
 181

<b>Table 1. Empirical tests of virulence evolution theory</b>	
<b>Key Finding</b>	<b>Key Empirical Evidence (Selected Papers)</b>
Virulence and transmission rate are positively correlated through replication rate	<i>Mus musculus</i> / <i>Plasmodium chabaudi</i> [30] ; <i>Homo sapiens</i> / <i>Plasmodium falciparum</i> [31] ; <i>Daphnia magna</i> / <i>Pasteuria ramosa</i> [32] ; <i>Homo sapiens</i> / HIV-1 [33] ; <i>Danaus plexippus</i> / <i>Ophryocystis elektroscirrha</i> [34] ; Meta-analysis of multiple systems [20]
Positive trait correlations saturate so that $R_0$ peaks at intermediate virulence	<i>Oryctolagus cuniculus</i> / Myxoma virus [17] (virulence-recovery rate); <i>Homo sapiens</i> / <i>Plasmodium falciparum</i> [31] (virulence-transmission rate) ; <i>Daphnia magna</i> / <i>Pasteuria ramosa</i> [32] (virulence rate-transmission rate) ; <i>Homo sapiens</i> / HIV-1 [33] (virulence rate-transmission rate), <i>Danaus plexippus</i> / <i>Ophryocystis elektroscirrha</i> [34] (virulence-transmission rate), <i>Gallus gallus domesticus</i> / Marek's disease virus [35] (virulence-transmission rate), <i>Haemorrhous mexicanu</i> / <i>Mycoplasma gallisepticum</i> [27] (virulence-transmission rate)
High susceptible density at the start of an epidemic selects for higher virulence	<i>Escherichia coli</i> / bacteriophage lambda [36]
Structured host populations select for less transmissible, prudent strategies	<i>Escherichia coli</i> / T4 coliphage [37] ; <i>Plodia interpunctella</i> / granulosis virus [38] ; <i>Escherichia coli</i> / bacteriophage lambda [39]
High virulence can trade-off with decreased host movement	<i>Danaus plexippus</i> / <i>Ophryocystis elektroscirrha</i> [40] ; <i>Haemorrhous mexicanu</i> / <i>Mycoplasma gallisepticum</i> [41] ; <i>Paramecium caudatum</i> / <i>Holospora undulata</i> [42]
Virulence evolves in natural epidemics of emerging disease	<i>Haemorrhous mexicanu</i> / <i>Mycoplasma gallisepticum</i> [43,44] (Less virulent strains spread fastest because of movement-virulence trade-offs and then are replaced by higher virulence strains. When hosts start evolving resistance, virulence continues to



	<p>increase through increased symptom severity rather than through replication rate)  <i>Oryctolagus cuniculus</i> / Myxoma virus [45] (Lower virulence quickly evolves from extremely high virulence introduction strains. When hosts start evolving resistance, virulence starts to increase)  <i>Corvus brachyrhynchos</i> / West Nile Virus [46] (A mutation conferring high virulence in American crows was positively selected, though this may have been a result of selection in another bird or vector species)</p>
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182

183 **4. What predicts virulence and transmission rate at spillover?**

184 **4(a). Virulence and transmission trade-offs act at spillover**

185 As we have outlined, theory on the virulence and transmission trade-off is based upon the  
 186 idea that pathogens will be selected towards an optimal level of virulence within the host  
 187 populations to which they are adapted [12]. Recently emerged zoonotic diseases do not have this  
 188 evolutionary history with human populations and are therefore highly unlikely to be at their  
 189 evolutionary optimum when they first emerge [22,47]. However, emerging pathogens may still be  
 190 regulated by an underlying virulence and transmission trade-off. In meta-analyses of recently  
 191 emerged viral zoonoses, excessively high virulence is associated with a lower  $R_0$  [22,48,49] and  
 192 this negative association supports the theoretical prediction that high virulence impedes pathogen  
 193 fitness. Theory also predicts a cost to excessively low virulence, an effect that is not supported  
 194 in these analyses [17,22]. However, this could easily result from discovery bias because we are  
 195 unlikely to notice low- $R_0$  zoonoses that cause only a few infections and have low virulence [11].  
 196 As such, there is little evidence to not expect emerging diseases to be governed by trade-offs  
 197 once they emerge into human populations.

198

199 **4(b). Virulence and transmission rates of zoonotic pathogens reflect evolutionary histories**  
 200 **with their reservoir hosts**

201 Emerging zoonoses vary widely in their virulence and transmission rates, but there are  
 202 key reservoir host characteristics that are associated with the pathogen's phenotype in humans  
 203 [22,48,50]. In particular, meta-analyses of recently emerged viral zoonoses have supported  
 204 phylogenetic trends in zoonotic potential [22]. The phylogenetic distance between a pathogen's  
 205 reservoir host and novel host predicts the pathogen's probability of being zoonotic [50], virulence  
 206 [22,51], and  $R_0$  [22,48]. Mammalian hosts closely related to humans (e.g. primates) harbor  
 207 zoonoses associated with lower human mortality and higher  $R_0$ , while more distantly related hosts

208 (most notably, bats) harbor highly virulent zoonoses that appear to be relatively maladapted for  
209 human-to-human transmission [22,52]. These phylogenetic trends can be understood if  
210 pathogens from distantly related reservoir hosts have evolved replication strategies adapted to  
211 their reservoir host's more dissimilar immunology, physiology, and ecology [22,47].

212 Importantly, these variations in pathogen virulence upon emergence reflect evolutionary  
213 histories within non-human reservoir hosts and demonstrate that emerging zoonotic diseases are  
214 not likely to be well adapted to human populations [22,47]. Reservoir host and pathogen traits can  
215 suggest what phenotypes a pathogen may have upon emergence, but do not tell us where these  
216 starting point phenotypes are relative to a pathogen's 'ideal' phenotypes in humans, since each  
217 pathogen will have a different evolutionary optimum depending on the nuances of its biology in  
218 the new host [9]. Because we cannot know where an emerging pathogen's starting point  
219 phenotypes are relative to its optimal phenotypes, we cannot precisely predict the direction of  
220 selection on virulence or transmission rate.

221

## 222 **5. Why is it difficult to predict how a novel zoonotic pathogen will evolve when it spills** 223 **over into humans?**

### 224 **5(a). Stochastic effects in small populations can overwhelm selection**

225 Because emerging zoonotic diseases are maladapted to human populations, we certainly  
226 expect for selection to favor improved pathogen fitness. However, this does not necessarily mean  
227 that pathogens will adaptively evolve [10,13]. A key tenant of evolutionary theory is that selection  
228 must act through a background of stochasticity and drift to result in adaptive evolution [53]. Small  
229 population sizes mean that both stochasticity and drift are relatively strong, and therefore the  
230 inevitably small population of infected individuals at the start of an epidemic means that  
231 stochasticity and drift are likely to overwhelm selection and determine the spread of mutants [53].  
232 Additionally, the existence of founder effects during epidemic range expansions results in spatial  
233 stochasticity analogous to genetic drift [54]. Thus, founder effects and variation in transmission  
234 due to host behavior and stochasticity likely determine the fate of mutants at the start of epidemics  
235 [10].

236 Additionally, adaptive evolution in acute, respiratory pathogens may be constrained by the  
237 small bottleneck sizes of transmission events [55]. Short infectious periods and small bottlenecks  
238 mean that it is less likely for a pathogen to have enough time within a host to generate adaptive  
239 mutations and select on those variants strongly enough for them to reach the high frequencies  
240 needed to transmit through tight bottlenecks [55]. This can impede adaptive evolution at the

241 population level [56]. All of these stochastic factors can overwhelm selection, especially at the  
242 start of an epidemic. However, as the population size of infected individuals increases or if there  
243 are mutations of large enough effect size, the balance between selection and stochasticity may  
244 shift towards selection and result in adaptive evolution.

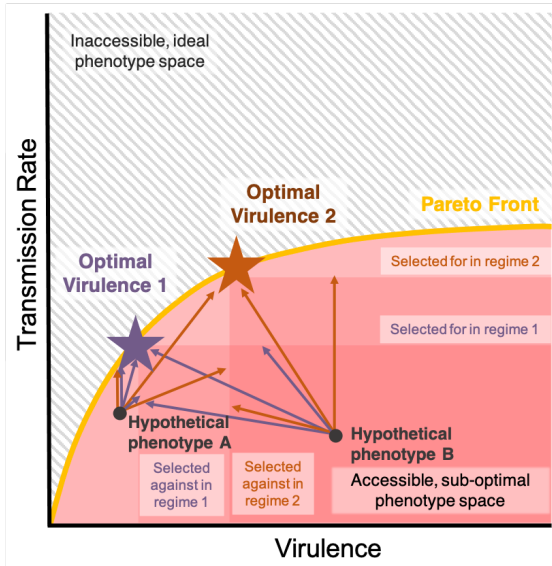
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#### 246 **5(b). Maladapted emerging zoonotic pathogens can evolve in unexpected ways**

247 There are many ways that emerging zoonotic pathogens can adapt to human hosts and  
248 the foremost is to improve their  $R_0$  [57]. Classic trade-off theory assumes that  $R_0$  should be  
249 maximized at intermediate virulence and transmission rates if these traits have tight, positive, and  
250 saturating correlations. However, these tight correlations assume that the pathogen is already  
251 relatively adapted to its host such that all potential adaptive mutations (for higher transmission  
252 rate or lower virulence) have costs (of higher virulence or lower transmission rate, respectively).  
253 This is unlikely to be the case for emerging zoonotic pathogens [22].

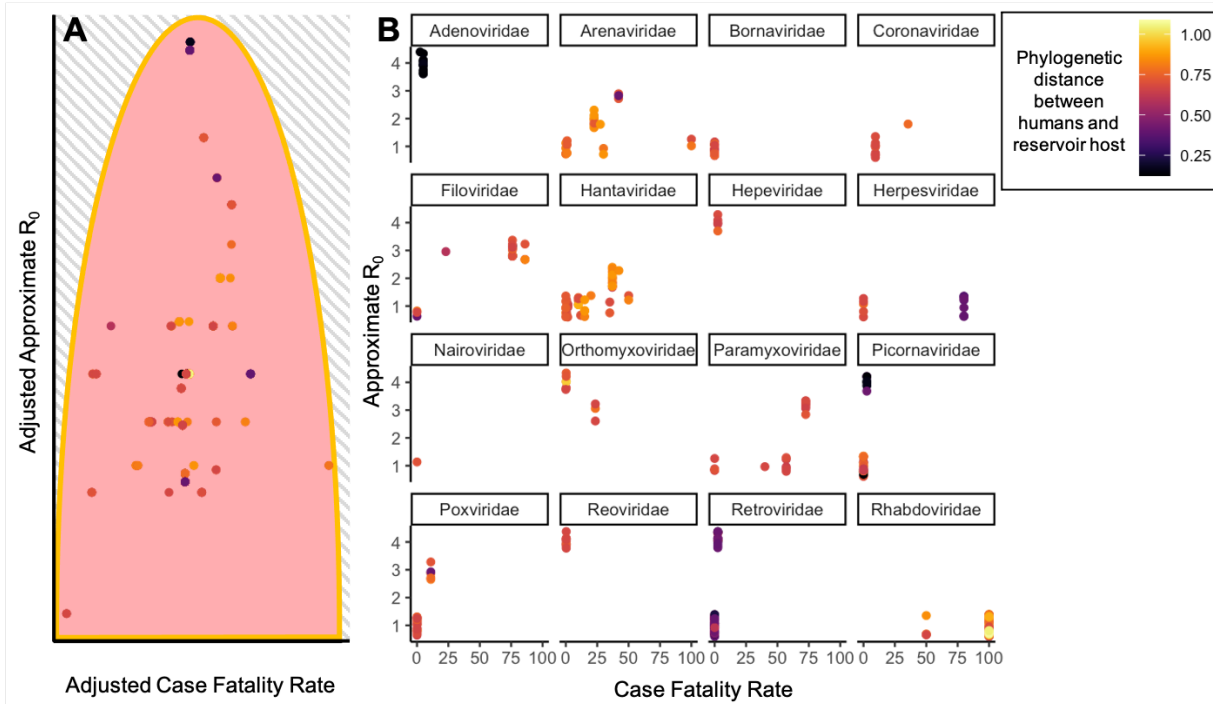
254 The concept of Pareto fronts describes scenarios where phenotypes can be in the region  
255 of sub-optimal phenotype space below the trade-off front (See Figure 3) [58]. The trade-off front  
256 (or Pareto front) separates these accessible, maladapted phenotype combinations from  
257 impossible, ideal phenotypes [58,59]. At the Pareto front, the two phenotypes trade-off with each  
258 other. Below the Pareto front, however, improvements in one trait may not affect the other trait as  
259 simple adaptations can be made before costs are incurred. Therefore, Pareto fronts determine  
260 which phenotype combinations are possible, and selection acts upon these possible phenotypes  
261 to move them towards more selectively advantageous regions.

262 Because they lack any evolutionary history with humans, emerging zoonotic diseases are  
263 unlikely have fixed all available 'no-cost' adaptations and thus likely have phenotypes below  
264 Pareto fronts (See Figure 4). Applied to virulence evolution, this means that zoonotic diseases  
265 emerging with lower than optimal transmission rates or higher than optimal virulence may initially  
266 select for no-cost improvements even if their 'optimal' phenotype is regulated by trade-offs (See  
267 Figure 3) [18]. This means that, in addition to not being able to precisely predict the direction of  
268 selection because we do not know where a pathogen's starting point phenotypes sit relative to  
269 their optimal phenotypes, we also cannot predict how any individual mutation improving  
270 transmission rate will affect virulence in a maladapted pathogen that starts below the Pareto front.



**Figure 3. Conceptual diagram of the Pareto front between virulence and transmission rate.** A Pareto front between virulence and transmission rate defines a region of accessible phenotype space. Theory determines where the ‘optimal strategy’ sits on the Pareto front to determine which regions of this phenotype space are selectively advantaged or disadvantaged. Phenotype combinations far from the Pareto front may technically be possible but would be highly selectively disadvantaged and likely to go extinct. Possible phenotypes can move towards their optimal strategy along any pathway within the accessible phenotype space. However, we cannot know where a hypothetical phenotype sits below its individual Pareto front. Selection for improved transmission rate can therefore involve decreases, no changes, or increases in virulence depending on the pathogen’s starting point and mutational availability.

271



272

273

**Figure 4. Recently emerged viral zoonoses loosely follow a Pareto front of virulence and  $R_0$  where  $R_0$  seems to be maximized at intermediate case fatality rates within viral families.** Data is from a dataset published in 2019 of recently emerged viral zoonoses from mammalian hosts [22]. Approximate  $R_0$  is classified from 1 (no human-to-human transmission) to 4 (endemic transmission). In figure 4A, dots represent plotted residuals from linear models of CFR and approximate  $R_0$  including virus family as a factor. By regressing out virus family, we somewhat control for the variation in trade-off shape for each virus and can make general observations across the dataset. Each dot therefore represents the virulence and  $R_0$  of an individual epidemic of viral zoonosis scaled by virus family. In figure 4B, CFR and Approximate  $R_0$  are directly plotted and separated by virus family so that the non-aggregated trends could be seen within virus families. In both panels, dots are colored by the phylogenetic distance between humans and the reservoir host. Plots were made with ‘ggplot2’. See supplement for code.

274 **6. How does a pathogen's optimal transmission rate and virulence depend on**  
275 **epidemiological characteristics and change over time?**

276 While we cannot predict exactly where the virulence and transmission rate of an emerging  
277 zoonotic disease sit relative to its Pareto front and thus also cannot predict whether fitness-  
278 improving mutations necessarily have costs, evolutionary epidemiology theory can tell us how  
279 different epidemiological characteristics shift which regions of the possible phenotype space are  
280 selectively advantageous. Additionally, while novel zoonotic pathogens sitting far below their  
281 Pareto front may initially have costless fitness-improving mutations, their evolution will be  
282 increasingly constrained by trade-offs as their fitness improves and they approach their Pareto  
283 front.

284 Thus, evolutionary epidemiology theory based upon the virulence and transmission trade-  
285 off can tell us what scenarios might select for different pathogen virulence and transmission rates.  
286 However, evolutionary epidemiology theory on the virulence and transmission trade-off is perhaps  
287 more nuanced than commonly appreciated. We've discussed how variations in trade-off shape  
288 can lead to different optimal phenotypes for different pathogens [12,17,21], but the optimal values  
289 of these rates can also depend on host and parasite epidemiological characteristics and change  
290 over time in an epidemic [4,12]. While saturating virulence and transmission rate trade-offs  
291 generally predict that intermediate virulence and transmission rate is optimal, certain  
292 epidemiological characteristics can bias a system towards selecting for higher transmission rate  
293 or less virulence depending on the relative selective importance of either trait. Below, we will  
294 discuss several bodies of theory that explore how different epidemiological characteristics effect  
295 optimal virulence and transmission rate, specifically focusing on those where the effect of the  
296 epidemiological characteristic being explored varies depending on the time scale of the epidemic.  
297 There are also several additional sections in the supplement on these effects in systems with  
298 multiple infection, environmental transmission ('curse of the pharaoh'), and antigenic escape  
299 (Supplemental Materials. S6(a), S6(b), S6(c), and Table S1).

300

301 **6(a). Selection favors high transmission rates when susceptible density is high at the start**  
302 **of an epidemic**

303 Classic models for virulence evolution examine long term evolutionary outcomes at  
304 equilibrium [60]. Selection on virulence and transmission rates during the start of an epidemic can  
305 be explored by using models that do not assume equilibrium [18,19,23,61,62]. These models  
306 allow for the existence of multiple simultaneous mutants so that the competitive fitness of each

307 can be assessed over shifting epidemiological conditions in time. They show that strains with  
308 higher transmission rates and virulence can be selected during epidemic growth stages, despite  
309  $R_0$  optimized (intermediate virulence) strains dominating at endemic equilibrium [19,23]. This is  
310 because strains with higher transmission rates spread fastest at the start of the epidemic when  
311 the density of susceptible hosts is high [19,23].

312 Intuitively, these results can be explained as: an infected host during the early stages of  
313 an epidemic encounters mostly susceptible hosts, so strains with higher transmission rates will  
314 have faster growth rates since they have shorter serial intervals (or infection generation times)  
315 than strains with higher  $R_0$  (but lower transmission rates) that produce more secondary infections  
316 over a longer infectious period but more slowly. For a simplified numeric example, a strain that  
317 has an infectious period of 2 days and infects 50% of its 2 contacts per day in an entirely  
318 susceptible population will only produce 2 new infections, but will double every 2 days.  
319 Comparatively, a strain that has an infectious period of 5 days and infects 40% of its 2 contacts  
320 per day in an entirely susceptible population will produce 4 new infections, but only double every  
321 2.5 days. Thus, the higher transmission rate strain can spread faster while susceptible host  
322 densities are high during epidemic growth stages, but the  $R_0$  optimized strain can outcompete it  
323 when susceptible density is low at endemic equilibrium because it produces a larger number of  
324 infections over its longer infectious period. Therefore, improvements in transmission rate are the  
325 most important at the start of an epidemic and can be selected for even if they have shorter  
326 infectious periods due to increased virulence. This also demonstrates that the high density of  
327 susceptible hosts early in epidemics crucially influences selection [12,18,19,23].

328

### 329 **6(c). Structured host populations select for prudent strategies at equilibrium, but** 330 **transiently select for virulent strategies at the epidemic front**

331 Classic virulence evolution trade-off theory assumes that transmission happens randomly  
332 in a homogeneously mixing population [12]. However, natural populations almost always have  
333 heterogeneous mixing patterns due to spatial structure and social networks [63,64]. In these  
334 structured populations, transmission occurs more often between neighboring individuals and  
335 those in social groups. This can lead to 'self-shading' where highly infectious strains rapidly  
336 deplete their local susceptible populations and compete for available hosts with related strains  
337 [63,65]. Thus, structured host populations select for lower pathogen infectivity and virulence at  
338 endemic equilibrium. However, the high availability of susceptible hosts at the start of an epidemic  
339 is likely to reduce the impact of self-shading and, moreover, pathogens need to have higher

340 transmission rates to seed an epidemic in a spatially structured population than in a well-mixed  
341 one [66]. Before equilibrium, the invasion front of a spatially structured epidemic also has a high  
342 local supply of susceptible hosts, which leads to a dynamic where virulent, high transmission rate  
343 strains are selected at the invasion front and then are succeeded by more prudent strategies as  
344 the local dynamics approach equilibrium [67,68]. Overall, then, it is possible that structure in host  
345 populations temporarily selects for higher virulence while the epidemic is spreading through  
346 mostly susceptible populations. However, if there are also trade-offs where high virulence  
347 impedes host movement, then the spatial front of the epidemic might instead have lower virulence  
348 [69]. As such, it is unclear how population structure and movement overall will select emerging  
349 pathogens during different parts of the epidemic.

350

351 **6(f). How might public health measures shape selection on virulence and transmission**  
352 **rate?**

353 The question of whether public health measures can purposely or inadvertently drive  
354 pathogen evolution naturally arises when discussing virulence evolution. Public health measures  
355 intentionally driving the evolution of virulence may be unrealistic in emerging zoonotic diseases  
356 because, as we have discussed, virulence evolution is very difficult to fully predict [9]. However,  
357 we can gain insight into how public health measures can inadvertently select on virulence. Non-  
358 pharmaceutical public health interventions for epidemics primarily aim to decrease transmission  
359 and therefore either stop the epidemic or slow it until vaccines and treatments can be developed.  
360 This decreases the total number of infected individuals, which will have the greatest impact on  
361 the total mortality burden of any epidemic [7]. This also limits the evolutionary potential of the  
362 pathogen by limiting the number of cases and therefore the strength of selection and opportunities  
363 for mutation [7]. However, some of these interventions may also contribute to the selection acting  
364 on the pathogen [7,9]. , decreased travel and extra-household contacts should alter the spatial  
365 and social structure of the population to make a more structured transmission network, which  
366 might prevent low transmission rate pathogens from spreading initially [63,66]. Second,  
367 quarantine of symptomatic individuals may select for decreased or altered symptoms, which could  
368 select for lower virulence if symptoms are linked to virulence [71]. Third, increased environmental  
369 sanitation decreases environmental transmission, thus potentially selecting for lower pathogen  
370 virulence under the ‘curse of the pharaoh’ hypothesis [70] (See Supplementary Material S6(b)).  
371 Finally, vaccines can sometimes create selection pressures on pathogens with potential  
372 evolutionary impacts to consider [72] (See Supplementary Material S6(c)).

373 While the most human mortality will be prevented by simply preventing transmission,  
374 considering the effects of control measures on pathogen evolution can, in principle, lead to better  
375 epidemic management [7]. Understanding host population characteristics creating strong  
376 selection for high transmission rate strategies could help distribute public health effort if there are  
377 limited resources [7]. However, a key point is that weak epidemic control measures that allow for  
378 extended transmission in humans increase the evolutionary potential of zoonotic pathogens  
379 because they allow for stronger selection and more mutations [7]. Thus, the best evolutionary  
380 management practice for an epidemic of a zoonotic infectious disease would be to suppress  
381 transmission using strong, rapid public health interventions.

382

## 383 **7. Conclusion**

384 In the face of the extraordinarily stressful circumstances of a global pandemic, we all  
385 understandably want simple answers for what will happen next and how the pathogen will evolve.  
386 Unfortunately, the simplest answer is that we cannot predict the evolution of any specific novel  
387 zoonotic pathogen. Its virulence and transmission rate may trade-off; it may be selected to  
388 increase its transmission rate; and the dynamics of selection may change with time.

389 The slightly more complicated answer is that, while we cannot predict how any specific  
390 pathogen will evolve, we do know how selection is expected to generally act on emerging zoonotic  
391 diseases and how different assumptions affect these predictions. We know that novel zoonotic  
392 pathogens emerge into the human population maladapted to human hosts [22,50]. Generally, we  
393 expect that virulence and transmission rate trade-off, leading to selection towards intermediate  
394 values of both [17]. However, we also know that a maladapted zoonotic pathogen's virulence and  
395 transmission phenotypes may start below the Pareto front, so selection for higher transmission  
396 rates can have decoupled effects on virulence [18]. Our theory also says that, with trade-offs, the  
397 optimal balance between virulence and transmission rate shifts depending on the time scale of  
398 the epidemic and different epidemiological and population characteristics [17,18].

399 All of these uncertainties make virulence evolution an academically interesting topic with  
400 a rich body of theory surrounding it, but no universal predictions [9]. Unfortunately, any sort of  
401 evolutionary prediction depends on a good understanding of how the phenotypes that the  
402 pathogen emerges with compare to their 'optimal' phenotypes in human populations; what fitness  
403 improving mutations the pathogen has available to it and what their associated trade-offs are; and  
404 how host population structure and epidemiological characteristics will shape the selection  
405 pressures on the pathogen. These data are exceptionally difficult to quickly gather. However,



406 despite our inability to conclusively predict how a pathogen will evolve, we do know that we can  
407 prevent it from doing so by implementing strong, rapid public health measures that suppress  
408 transmission early on since this will decrease the evolutionary potential of such pathogens while  
409 also decreasing the total mortality burden by limiting the number of people infected.

410

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415

#### 416 **Author Contributions**

417 All authors researched and edited the paper. EV and MB conceptualized and wrote the paper.

418

#### 419 **Data Availability**

420 No novel data is used in this manuscript; data used is publicly available as online Supplementary  
421 Material from [22]. The annotated R script used for analysis is provided in the supplement.

422

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429

#### 430 **Competing Interests**

431 We declare no competing interests.

432

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