

1 **The expected effect of deleterious mutations on within-host adaptation of**
2 **pathogens**

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22

23

24 **ABSTRACT**

25 Adaptation is a common theme in both pathogen emergence, for example in
26 zoonotic cross-species transmission, and pathogen control, where adaptation
27 might limit the effect of the immune response and antiviral treatment. When such
28 evolution requires deleterious intermediate mutations, fitness ridges and valleys
29 arise in the pathogen's fitness landscape. The effect of deleterious intermediate
30 mutations on within-host pathogen adaptation is examined with deterministic
31 calculations, appropriate for pathogens replicating in large populations with high
32 error rates. The effect of deleterious intermediates on pathogen adaptation is
33 smaller than their name might suggest: when two mutations are required, and
34 each individual single mutation is fully deleterious, the pathogen can jump across
35 the fitness valley by obtaining two mutations at once, leading to a proportion of
36 adapted mutant that is 20-fold lower than for the situation where all mutants are
37 neutral. The negative effects of deleterious intermediates are typically
38 substantially smaller, and outweighed, by fitness advantages of the adapted
39 mutant. Moreover, requiring a specific mutation order has a substantially smaller
40 effect on pathogen adaptation than the effect of all intermediates being
41 deleterious. These results can be rationalized when calculating the number of
42 routes of mutation available to the pathogen, providing a simple approach to
43 estimate the effect of deleterious mutations. The calculations discussed here are
44 applicable when assessing the effect of deleterious mutations on the within-host
45 adaptation of pathogens, for example in the context of zoonotic emergence,
46 antigenic escape, and drug resistance.

47 **IMPORTANCE**

48 Adaptation is critical for pathogens after zoonotic transmission into a new host
49 species, or to achieve antigenic immune escape and drug resistance. Using a
50 deterministic approach, the effects of deleterious intermediate mutations on
51 pathogen adaptation are calculated whilst avoiding commonly made
52 simplifications that do not apply to large pathogen populations replicating with
53 high mutations rates. Perhaps unexpectedly, pathogen adaptation does not halt
54 when the intermediate mutations are fully deleterious. Negative effects of
55 deleterious mutations are substantially outweighed by fitness gains of adaptation.
56 To gain an understanding of the effect of deleterious mutations on pathogen
57 adaptation, a simple approach is introduced that counts the number of routes
58 available to the pathogen with and without deleterious intermediate mutations.
59 This methodology enables a straightforward calculation of the proportion of the
60 pathogen population that will cross a fitness valley or traverse a fitness ridge,
61 without reverting to more complicated mathematical models.

62

63 **INTRODUCTION**

64

65 The fitness landscape of a pathogen is likely to have a rugged shape and consist
66 of multiple optima. Reductions in fitness occur when underlying combinations of
67 genetic mutations lead to an unfit or deleterious phenotype, creating depressions
68 in the fitness landscape. One phenomenon causing sharp peaks and troughs in
69 the fitness landscape is sign epistasis, where a beneficial adaptation involves a
70 combination of individually deleterious mutations (1–5). In the case where
71 intermediate mutations are less fit than the wild type and adapted virus, a fitness
72 valley is created – a barrier of disadvantageous mutations hampering the access
73 to other landscape regions (4, 6). If there is a specific order in which mutations
74 can occur without compromising the fitness, for example where compensatory or
75 obligatory co-mutations can remove the deleterious effect of another mutation,
76 the landscape contains a fitness ridge. Such fitness valleys and ridges are
77 commonplace in virology, as will be illustrated with examples drawn from the
78 influenza field.

79

80 During zoonotic overspill infections of an avian influenza virus into humans,
81 pressure exists for the pathogen to adapt to this possible new host (7–9). The
82 virus was fit in its original host, and needs to be fit in the new host, but this
83 adaptation process might require deleterious intermediate mutations. The need
84 for adaptation of a zoonotic pathogen is illustrated by the requirement of a
85 combination of mutations in avian A/H5N1 virus for airborne transmission

86 between mammals (10, 11). Interestingly, two of the mutations that were found
87 necessary to confer airborne transmissibility, polymerase basic protein 2 (PB2)
88 E627K and polymerase basic protein 1 (PB1) H99Y (11, 12), increased the
89 fitness of the adapted virus if both mutations occurred together, as inferred from
90 substantially larger plaque sizes than the wild type, yet each individual mutation
91 decreased the fitness compared to the wild type virus (12). Similarly, Imai *et al.*
92 showed that the receptor-binding mutations N224K and Q226L in the
93 hemagglutinin gene (HA), required for an airborne transmissible phenotype of
94 A/H5N1, reduced the stability of HA, but could be compensated for by mutation
95 T318I in the HA stalk, which restored protein stability (10). Although the
96 stabilizing mutation was not essential for virus survival, it did substantially
97 increase viral fitness.

98

99 Another example of deleterious intermediate mutations is escape from pre-
100 existing host immunity through fitness-decreasing mutations for antigenically
101 variable pathogens (13, 14). For example, the altered receptor-binding avidity
102 and lower replication resulting from the antigenic escape mutation HA K165E in
103 A/Puerto Rico/8/1934(H1N1) could be compensated for by mutations in HA or
104 the neuraminidase (NA) (15, 16), and stabilizing mutations were required to
105 occur prior to the introduction of immune-escape mutations in influenza A/H3N2
106 virus (17). Similarly, there are numerous examples where antiviral-resistance
107 conferring mutations come at a fitness cost for the virus, that can be
108 compensated for by other mutations: several neuraminidase substitutions can

109 occur and have occurred as either permissive or compensatory mutations to
110 counteract the adverse fitness effects of the oseltamivir-resistance mutation NA
111 H275Y in influenza A/H1N1 virus (18–20); and similarly the I222V NA mutation in
112 A/H3N2 partially restored the viral fitness-decreasing oseltamivir-resistance
113 mutation NA E119V (21).

114

115 The name “deleterious mutation” may suggest that the existence of such
116 mutations is unlikely, and thus the expectation that the crossing of a fitness valley
117 comprised of individually deleterious mutations is difficult, if not impossible.
118 Indeed, when evolution is described as an adaptive walk or directed evolution,
119 adaptation consists of a series of incrementally neutral or beneficial mutations,
120 and thus the crossing of the fitness valley would be technically impossible (22–
121 26). Also the possibility of obtaining several mutations at once, to “jump across” a
122 fitness valley, is not considered in some theoretical models (27–29). A
123 methodological framework frequently used to study pathogen evolution making
124 such assumptions is the “strong-selection-weak-mutation” (SSWM)
125 approximation (30, 31). Models using the SSWM assumption describe the
126 evolutionary trajectory of a population where selective sweeps cause the
127 sequential fixation of advantageous mutations, whilst deleterious or neutral
128 mutations are disregarded (32–34).

129

130 Here, we demonstrate how pathogens replicating in large population sizes and
131 with high error rates, such as RNA viruses, can cross fitness valleys, based on a

132 simple, and appropriate, deterministic model of within-host pathogen evolution.
133 Instead of following the evolution of a pathogen population toward fixation of
134 certain mutations, as is for example done in SSWM models, we calculate the
135 probability of a randomly drawn virion from the within-host virus population after
136 initial infection with a single genotype to have obtained a set of mutations after a
137 given number of replication rounds. This probability, when multiplied with the
138 pathogen population size, gives the expected number of virions with this specific
139 set of mutations. In other words, the probability is directly related to the
140 proportion of viruses in the total within-host population with this set of mutations.

141

142 This probability of a virion to have a set of mutations is highly relevant, because
143 increased proportions are likely to correspond to increased probability of spread
144 of such mutants. When the bottleneck of transmission is narrow, for example in
145 the case where a single virion is transmitted to the next host, the probability
146 describes the likelihood that infection of the next host will begin with the adapted
147 virus. Alternatively, if the bottleneck is wide, the expected proportion of adapted
148 virus at the start of the infection of the next host can be calculated and used to
149 estimate the chances of further adaptation.

150

151 In this manuscript, we calculate the effects of deleterious mutations, fitness
152 valleys and fitness ridges on within-host pathogen evolution using a
153 straightforward deterministic model (35, 36). Such a deterministic probability
154 calculation is appropriate for studying the dynamics and evolution of large

155 populations with asexual reproduction at a high mutation rate, such as most RNA
156 viruses, because stochastic effects play a limited role. The proportion of the
157 adapted mutant is calculated for varying valley depths (i.e. the fitness of the
158 deleterious mutation) and breadths (i.e. number of deleterious mutations). Finally,
159 we also describe the probability of traversing a fitness ridge, for varying numbers
160 of mutations that need to be acquired in order.

161

162 **METHODS**

163

164 The within-host population dynamics of virus mutants were calculated as
165 deterministic probabilities, based on the methodology described previously (35,
166 36). In this calculation, the errors made by the virus polymerase are represented
167 by an error rate, and form the source of introduction of mutations, but the
168 approach can equally be used for non-viral pathogens, where mutations are
169 introduced through another mechanism. The probability of accumulating
170 mutations and the within-host evolutionary dynamics of the virus population are
171 explored as a function of the fitness of the wild type, intermediate and adapted
172 mutants.

173

174 **Calculating virus populations**

175 A virus type j is a virus with a particular set of mutations. The probability of each
176 virus type (N_j) after a replication round is given by the sum of contributions from
177 each type in the previous replication rounds:

178

179 **eq. 1**
$$N_j(t) = \sum_i [N_i(t-1)\mu_{ij}]$$

180

181 Where μ_{ij} is the probability of type i mutating to type j , and each type contributes
182 exactly its expected value. If the mutation rate μ_{ij} is low, the main contribution to
183 the proportion of the population that is N_j at time t will be from the proportion of
184 the population that was N_j at time $t-1$, and a smaller contribution from virus type

185 N_i at time $t-1$ that mutated into type N_j . The probability of mutation μ is calculated
186 as follows:

187

188 **eq. 2**
$$\mu_{ij} = \prod_{\{m=0\}}(1 - r) \prod_{\{m=1\}} r$$

189

190 Where r is the polymerase error rate. Thus, μ_{ij} is the product of the probabilities
191 of non-mutation ($1-r$) for the set $\{m=0\}$, i.e. positions for which no mutation is
192 required, and of the probability for mutation (r) for the set of positions that need to
193 mutate $\{m=1\}$.

194

195 **Accounting for fitness values**

196 The deleterious and beneficial selection values were incorporated by adjusting
197 the “progeny” of each virion to express the fitness disadvantage or advantage in
198 each genome replication step, prior to the start of the next generation. The
199 starting population (generation zero) consists only of zero-mutant, the starting
200 virus. After the first replication round (*equation 1*), the population of each type N_i
201 is multiplied by its relative fitness f_i , and the population is normalized (such that \sum_i
202 $N_{i_adj} = 1$) through division by the sum of the fitness-weighted prevalence all
203 types:

204

205 **eq. 3**
$$N_{i_adj}(t) = \frac{N_i(t) * f_i}{\sum_i [N_i(t) * f_i]}$$

206

207 The N_{i_adj} represent the populations at the start of the next genome replication
208 step, and are used as N_i in *equation 1* in the multiplication with the mutation
209 matrix. When calculating the effect of deleterious mutations the fitness f is varied,
210 and a “fully deleterious mutant” has a fitness of $f = 0$, which causes relative
211 increases in the probability of the other virus types in the total virus population.

212

213 Unless otherwise noted, the fitness of the wild type (i.e. starting) virus, and the
214 final type with the full set of mutations of interest, the “adapted virus”, is neutral: f
215 = 1.

216

217 Because this model normalizes the virus population via *equation 3*, and accounts
218 for back-mutations in *equation 2*, the results are slightly different from the short-
219 hand formula introduced in *equation 4* in the results section. For this reason,
220 calculating the number of routes (see below) is a fast and informative approach
221 to very closely approximate the probability of a certain set of mutations, but is not
222 analytically identical to the modeling results.

223

224 **Stochastic model runs**

225 In addition to the deterministic modeling results above, a set of stochastic
226 discrete-time multi-type branching process simulations were run, see also
227 Russell *et al.* (35). The starting population of a single virion expanded
228 exponentially with a branching factor of 32 (leading to 10^3 virions produced per
229 infected cell, after the two genome replication steps), until the population size

230 exceeded 10^{10} virions, from which point onwards the branching factor was set to
231 1. For each genome replication step, the expected number of each mutant type
232 was determined with a Poisson distributed random variable with the expectation
233 value based on the mutation matrix shown in *equation 2* and the number of
234 virions of each mutant type existing before the replication step, as was done for
235 *equation 1*. This number was then multiplied by the relative fitness of each type,
236 and rounded to the nearest integer, prior to starting the next genome replication
237 step. We performed 10,000 stochastic runs for each of the 101 settings of fitness
238 of the deleterious intermediate mutants (between 0 and 1 in steps of 0.01). The
239 intensity of the shade of the pale red and blue colors is calculated based on the
240 \log_2 of the number simulation runs that have the resulting proportion of double
241 mutants for each fitness setting; the average proportion across the 10,000 runs
242 per fitness setting is indicated with the line connecting the circles.

243

244 **Determining the route**

245 We introduce the terminology “through singles” to mean the process by which the
246 two mutations are acquired through separate single mutations occurring in
247 distinct replication rounds, and “through doubles” to mean to process where two
248 mutations are achieved by mutating both sites in a single replication round. We
249 investigated the probability of a double mutant to occur through doubles by
250 setting the μ for single mutations to zero in the mutation matrix. The through
251 singles probability is calculated as the difference between the probability when all
252 routes are allowed, and the through doubles probability.

253

254 Similarly, to calculate how often the required set of mutations was achieved
255 through a specific order, the fitness of any non-order mutant was set to zero. The
256 difference between the probability calculated if any order is allowed and the
257 probability when only a specific order is available determines the probability of
258 non-order mutation routes.

259

260 The fraction of available routes is calculated as the number of available routes
261 given the constraints divided by the number of original routes.

262

263 **Parameter choice**

264 The mutation rate is parameterized by the current best estimate for the influenza
265 virus polymerase error rate ($r = 1 \times 10^{-5}$ mutations per site, per genome replication
266 (37, 38)), and can trivially be adjusted for other mutation rates – indeed all results
267 in the manuscript are *not* specific to influenza virus, or viruses in general, but to
268 all large populations where mutations occur.

269

270 A “replication round” in this manuscript refers to any step in which RNA is
271 synthesized, because in each round of replication polymerase errors can be
272 introduced. For influenza viruses, where vRNA is replicated into cRNA and then
273 cRNA is copied into vRNA, there are (at least) two replication rounds per cell
274 cycle. Results are shown after 20 viral replication rounds, which corresponds to
275 five days of influenza virus infection (where each replication round lasts around 6

276 hours, and virions exit the cell after 12 hours), but again, the number of
277 replication rounds can be varied in the equations above.

278 **RESULTS**

279 **Adaptation depends on the fitness of the deleterious intermediate**
280 **mutations**

281

282 When all mutations are neutral, a simple probabilistic calculation of mutation
283 accumulation closely approximates the probability that any randomly drawn virion
284 from the within-host virus population in an individual initially infected with a single
285 genotype, would have mutated the m sites of interest over time (the number of
286 replication rounds, t):

287

288 **eq. 4** $p(m,t) = t^m r^m$

289

290 This equation multiplies the probability of getting m mutations (based on the
291 polymerase error rate r), r^m , with the number of combinatorial options to acquire
292 these m mutations over t generations (t^m), see also Russell *et al.* (35) and
293 Gokhale *et al.* (39).

294

295 The probability that a given virion will have mutated $m = 2$ sites after $t = 20$
296 replication rounds with a polymerase error rate $r = 1 \times 10^{-5}$ is approximated by
297 *equation 4* as 4×10^{-8} . Naturally, this probability of observing both sites mutated
298 will be less if either of the individual mutations is deleterious. If both individual
299 mutations are deleterious, the pathogen will have to get across a fitness valley.
300 The fitness of each single deleterious mutant determines the likelihood of the

301 virion to cross this fitness valley. A fully deleterious mutation has a relative fitness
302 of 0, which means that no progeny is made from these virions at all, whilst for a
303 relative fitness of 0.5 half as much progeny descends from these virions
304 compared to virions with a relative fitness of 1.

305

306 *Figure 1A* explores how the probability of a pathogen to cross a fitness valley
307 depends on the deleterious effect of the intermediate mutations. In this scenario,
308 each individual mutation is equally deleterious, and the wild type (starting) and
309 the virus with the two required mutations (the “adapted virus”) have neutral
310 fitness. The blue line shows the deterministic probability of a virion to be a double
311 mutant as a function of the fitness of the intermediate mutants. In the neutral
312 scenario without any fitness valley, where the fitness of each intermediate mutant
313 is 1, the probability of the double mutant after 20 replication rounds is, as
314 approximated above, 4×10^{-8} . As the relative fitness of each intermediate mutant
315 decreases toward zero (fully deleterious), the probability that any random virion
316 in the virus mixture is a double mutant decreases to 2×10^{-9} . Note that, despite
317 the two intermediate mutants being fully deleterious, the probability of the double
318 mutant is only twenty-fold lower than without the fully deleterious fitness valley.
319 The pale region is composed of 10,000 stochastic model simulations for each of
320 the 101 different settings of fitness f . The average of these runs, indicated by the
321 connected circles, is somewhat lower than the deterministic calculations, most
322 visibly for intermediate values of deleterious fitness. These stochastic simulations

323 highlight that even though stochastic effects may play a role, double mutations
324 do occur regularly, even when the intermediate mutants are fully deleterious.

325

326 Such double mutants can arise when both mutations were acquired
327 simultaneously in a single replication round, the “through doubles” mechanism.

328 The purple line in *Figure 1B* shows how much this mechanism of acquiring both
329 mutations at once contributes toward the likelihood of a virion being a double
330 mutant. Note that this probability is not affected by the relative fitness: because
331 the deleterious intermediates were never formed when two mutations were
332 obtained at once, the virions avoid having to incur the designated fitness cost.

333 The fitness valley is not crossed, but the virus “jumps” over it. The cyan line
334 describes the alternative “through singles” mechanism where the two single
335 mutations were obtained in distinct replications rounds – the situation in which
336 the virions did incur the deleterious cost of the intermediate, and actually crossed
337 through the valley. In *Figure 1B*, it can be seen that this contribution depends
338 strongly on the fitness of the intermediate single mutants. If the intermediate
339 mutants are neutral, or have a high relative fitness, the through singles
340 mechanism is the main contributor toward the probability of acquiring a double
341 mutant (right hand side of *Figure 1B*). However, when the intermediate single
342 mutants are highly deleterious the main contributor to the probability of a double
343 mutant is the through doubles mechanism.

344

345 Returning to *Figure 1A*, it appears that the through doubles mechanism is less
346 sensitive to stochastic variations than the through singles mechanism, and the
347 deviation between the deterministic model and stochastic results is largest for
348 intermediate values of deleterious fitness. Here, the non-negligible deleterious
349 cost causes stochastic loss of single mutants before the second mutation occurs.
350 Such stochastic losses are less prominent for fit intermediates ($f = 1$) or identical
351 or similar to the losses calculated in the deterministic model for highly deleterious
352 intermediates.

353

354 **An intuitive understanding: counting the number of “routes”**

355 Although the “through doubles” and “through singles” mechanisms in *Figure 1*
356 both require two polymerase errors, the probability of which is r^2 , the relative
357 contribution of through singles to obtain two mutations is larger than through
358 doubles at $f = 1$. This phenomenon can be understood by considering “the
359 number of routes”. The through doubles route can happen once in each
360 replication round, and thus in t different ways (here 20). However, to get two
361 single mutations, there can be e.g. single mutations in two subsequent rounds
362 (for which there are 19×2 options – the factor of two accounts for which of the
363 mutations is first), or single mutations in two replication rounds separated by a
364 replication round without mutation (for which there are 18×2 options), and so on,
365 until there is one single mutation in round 1 and one in round 20 (for which there
366 are 2 options only). The sum of these possibilities is 380 routes, which when
367 combined with the 20 routes of through doubles, corresponds to 20^2 ways to

368 obtain two mutations in twenty replication rounds, i.e. the factor t^m in *equation 4*.
369 Although the term t^m is nothing more than a combinatorial factor, it was found
370 that explicitly analyzing the number of routes represented by this term is useful
371 for reasoning about the expected effects of varying fitness valley shapes.

372

373 If the single mutants are fully deleterious, the effective number of routes to obtain
374 a double mutant through singles is 0, because a double mutant can never arise
375 from a single mutant if single mutants do not have progeny. In this situation, only
376 the through doubles mechanism is possible, and thus 20 out of the original 400
377 routes remain, causing a reduction in the probability of a double mutant by a
378 factor 0.05 (the probability decreased from 4×10^{-8} to 2×10^{-9}).

379

380 A general description to calculate the effective number of routes to obtain a
381 double mutant through singles for any relative fitness f of the single mutations
382 can be given as well: if single mutations happen in successive replication rounds
383 ($delay = 1$), the fitness cost is incurred once, if they are separated by one
384 replication round ($delay = 2$), the fitness cost is incurred twice, whilst if the single
385 mutations are 19 generations apart ($delay = 19$), the fitness cost f is incurred 19
386 times. In total, the effective contribution to the number of routes through singles
387 weighted by the incurred deleterious cost is given by:

388

389 **eq. 5** Effective number of routes $= \sum_{delay=1}^{delay=(t-1)} [2 * (t - delay) * f^{delay}]$

390

391 Where t is the number of replication rounds, as before, and $delay$ is the time
392 lapse in replication rounds between the two single mutations (for $t = 20$
393 replication rounds, the maximum $delay$ is 19). The factor of two reflects the fact
394 that the single mutations can be acquired in two different orders; the term $(t -$
395 $delay)$ represents the number of options for any given delay (19 for a delay of 1,
396 18 for a delay of 2, etc.); while f^{delay} is the penalty term for the incurred fitness
397 cost over $delay$ rounds of replication.

398

399 *Figure 2* illustrates how the number of effective routes is composed of the
400 contribution of the through doubles mechanism (in grey), and the different single-
401 single mutation routes. If the relative fitness of the intermediate mutants is 0, the
402 through doubles mechanism is the only contributor to the number of routes, as
403 was seen in *Figure 1*. Again, the effective number of routes for the through
404 doubles mechanism is independent of the fitness of the deleterious single
405 mutants, as any deleterious fitness cost is not incurred. *Figure 2* also
406 demonstrates that the effective number of routes of two single mutations
407 separated by many replication rounds (e.g. $delay = 19$, in pink) is substantially
408 smaller than the effective number of routes for two subsequent single mutations
409 ($delay = 1$, in red). The reason for this is twofold: first, if there are 20 replication
410 rounds, there are 19×2 routes to generate two single mutations 1 generation
411 apart, whilst there are only 1×2 routes to obtain two single mutations 19
412 generations apart. Second, any deleterious effect of the single mutants is
413 incurred for more replication rounds if the delay between the two single mutations

414 is longer, and thus the contribution of these single-single routes with longer
415 delays decreases even more as the intermediate mutants become more
416 deleterious.

417

418 *Table 1* shows that the total number of routes increases (t^m) as more mutations
419 are required, listing the results for two to seven mutations required. It also shows
420 the fraction of routes remaining when all intermediate mutants are fully
421 deleterious. In the case where five mutations are required, for example, this
422 means that all individual and combined intermediates (and thus all single, double,
423 triple and quadruple mutants) are deleterious. Because all intermediate mutants
424 are fully deleterious, all mutations have to be acquired at once, for which there
425 are t options: so t out of t^m routes remain. Although there were initially many
426 routes to acquire 5 mutations (20^5), only 20 remain.

427

428 When comparing the fraction of available routes for the situation where 5
429 mutations are required, 6.3×10^{-6} , with the fraction when 2 mutations are required,
430 5.0×10^{-2} , it is clear that the fraction of the available number of routes decreases
431 greatly as the number of intermediate deleterious mutations increases. Note, in
432 addition to a larger number of deleterious intermediate mutants slowing down the
433 viral adaptation, there is also the increased difficulty of acquiring more mutations
434 in the first place (which is given by r^m).

435

436 When a set of mutations is required of which only some are deleterious, the ratio
437 of the effective number of routes compared to the total number of routes when
438 that subset of mutations was not deleterious is the same as the fraction of routes
439 available for the number of deleterious mutations. As an example, consider the
440 situation where 5 mutations are required, and two of the mutations need to be
441 acquired as a double. When none of the 5 mutations are deleterious, there are
442 20^5 routes (t^m). When the two mutations are individually fully deleterious, the
443 second mutation of the double pair needs to occur simultaneously with the first
444 mutation of the pair. As a result, the timing of mutation, for which there are 20
445 options if there are 20 replication rounds, needs only to be established for 4
446 mutations, as the timing of the last mutation needs to be identical to the timing of
447 the other mutation in the pair. Hence, when two of the five mutations need to be
448 acquired as a double, there are 20^4 routes left, and the fraction $20^4/20^5$ is 0.05,
449 see 2 mutations required in *Table 1*.

450

451 **The effect of deleterious intermediates is outweighed by the fitness**
452 **advantage of adaptation**

453 In the previous calculations, we studied situations where the fully adapted mutant
454 had neutral fitness, and the number of available routes could directly be used
455 when calculating the probability of a virion being a fully adapted mutant. Next, we
456 investigate whether the deleterious cost of an intermediate mutant can be
457 outweighed by the fitness gain that would be obtained upon achieving the full set
458 of mutations, for example as a result of obtaining a certain beneficial phenotype

459 such as antigenic escape or increased replication. In *Figure 3*, the probability of
460 any virion being a double mutant after $t = 20$ replication rounds is indicated by
461 color, as a function of the relative fitness advantage f of the double mutant,
462 varied from 1 to 4, and the relative fitness of the deleterious single mutants,
463 varied from 0 to 1.

464

465 *Figure 3* demonstrates that the probability of a random virion having obtained two
466 mutations after 20 replication rounds varied relatively little with the fitness of the
467 deleterious intermediates (along the x-axis): for example, the maximum change
468 in the neutral scenario (see *Figure 1*) was 20-fold, which corresponds to 1.3 units
469 on a \log_{10} axis. In contrast, the fitness gain of the double mutant causes changes
470 across 9 orders of magnitude, and this fitness gain therefore appears to be the
471 main determinant of the probability of a double mutant. The mechanism behind
472 these observations is that, in contrast to the deleterious cost, which is often
473 incurred only briefly, or avoided altogether by obtaining both mutations at once,
474 the fitness benefit of the double mutant is incurred in every single replication
475 round once it has arisen, hence exponentially increasing its presence in the
476 pathogen population. For fitness $f = 4$ of the double mutant, the minimum
477 probability of a virion to be a double mutant, across all fitnesses of the single
478 mutants, was 0.99998.

479

480 **Adaptation via fitness ridges: compensating mutations imposing**
481 **order**

482 Requiring mutations to occur in a specific order is a special case of deleterious
483 mutations: imagine the scenario in which two mutations are required, whereby
484 one single mutation α compensates for or removes the deleterious effect of the
485 other single mutation β (both the single intermediate α and the double mutant $\alpha\beta$
486 have neutral fitness). To understand the effect of such imposed order on the
487 probability of obtaining a certain mutant, again the number of effective routes
488 calculation is helpful.

489

490 In the scenario where the double mutant and mutation α are neutral, while
491 mutation of only site β is fully deleterious, the two ways toward the double mutant
492 are to either get both mutations simultaneously, or to obtain the non-deleterious,
493 compensating mutation α before mutating the site β . As explained above, there
494 are 20 routes out of 400 to obtain both mutations at once, and half of the
495 remaining 380 routes will have had the compensating mutation α prior to
496 mutation of site β : in total 210 out of the 400 routes remain. Even though
497 mutation β was fully deleterious, 52.5% of the routes are still available, thus
498 incurring only a 2-fold reduction in the total proportion of double mutant.

499

500 *Table 2* shows the number of allowed routes when all mutations need to be
501 obtained in order, for situations where two to seven mutations are required. As
502 more order is required, the reduction in fraction of allowed routes increases: for

503 stringent ordering of five mutations, only 1.3% of the routes remain, which is less
504 than the 52.5% for requiring order of two mutations.

505

506 Because imposing order does not necessarily require multiple mutations to occur
507 at once, the fraction of available routes is substantially larger in the situation
508 where order is required than for the situation where all intermediate mutants were
509 deleterious. For example, when requiring five mutations, 1.3% of the routes
510 remain if requiring specific ordering of these five mutations, whereas only
511 0.00063% of the routes remained when all intermediates were deleterious (see
512 *Table 1*). Thus, using available fitness ridges is always, and often considerably,
513 easier than jumping across or crossing a fitness valley.

514

515 The red line in *Figure 4A* also shows that the effect of requiring order on the
516 probability of a virion obtaining two mutations is relatively small, even if the non-
517 ordered single mutant is fully deleterious (compare $f = 0$ and $f = 1$), especially
518 when compared to the situation where both single mutants were deleterious (blue
519 line). The pale region is composed of 10,000 stochastic model simulations for the
520 fitness ridge, in red, and fitness valley, in blue. The average of the stochastic
521 runs, shown as circles, again indicate how traversing via a fitness ridge is
522 substantially more likely than jumping or crossing a fitness valley. Moreover,
523 stochastic effects play virtually no role for the outcome of a virion that can travel
524 via a fitness ridge, as the results are very similar to the deterministic model.

525

526 *Figure 4B* shows the contribution of the three different mechanisms that could
527 lead to a double mutant. First, the virus could follow the imposed order and travel
528 via two subsequent mutations along the fitness ridge. Second, the virus could
529 simultaneously mutate both sites and jump across the surrounding fitness valley.
530 Both of these mechanisms do not violate the imposed order, and their
531 contributions in *Figure 4B* are independent of the fitness of the deleterious single
532 mutant. Third, the virus could disobey the imposed order and obtain the non-
533 ordered single mutation first, which will incur the deleterious cost for a certain
534 time. The contribution of this latter mechanism depends on the fitness of the non-
535 ordered deleterious single mutant, and becomes zero when $f = 0$. As f decreases,
536 the ridge in the fitness valley becomes the main mechanism toward obtaining the
537 set of mutations. In general, following the imposed order and travelling via the
538 fitness ridge becomes more attractive as the fitness valley deepens and widens,
539 as non-ordered single intermediates become even less viable, and obtaining
540 simultaneous mutations even less likely.
541

542 **DISCUSSION**

543

544 Using probabilistic calculations on within-host genetic evolution, we found that
545 the effect of a fitness valley of deleterious intermediate mutations on adaptation
546 is much smaller than might be expected, and that the effect of requiring a specific
547 order for mutations to occur is even smaller. In coinfecting individuals,
548 mechanisms such as reassortment (if the mutations of interest are on separate
549 genes) and recombination are additionally affecting the ability of viruses to
550 overcome fitness valleys, processes that are not currently included in the model.
551 Instead, we calculated, based on within-host evolution of a single starting virus
552 genotype infecting an individual, the probability of any virion getting a set of host-
553 adaptation mutations. This probability is directly related to the expected
554 proportion of adapted mutant in the total population. The equations and
555 calculations presented in this work can be used in any situation with fitness
556 valleys and ridges where the deterministic assumptions are fulfilled, and the
557 population reproduces asexually. As a result, this method can aid the study on
558 the effects of deleterious mutations in a wide range of pathogens, including, for
559 example, tuberculosis and HIV (5, 40, 41).

560

561 The methodology of counting the number of routes is a straightforward approach
562 to calculate the effect of deleterious intermediate mutants, and understand the
563 ways by which pathogen populations traverse fitness valleys and ridges. For
564 example, in the situation where a virus requires two mutations that are each

565 individually fully deleterious, the evolution is not halted, as this trap is avoided by
566 acquiring multiple mutations at once. In addition, if viruses need to follow a
567 specific order of mutation, the out-of-order intermediate mutants can be
568 described as fully deleterious. If only a handful of mutations need to be acquired
569 in order, the influence on the adaptation of the virus is minimal, because the
570 compensating mutations will occur beforehand without much difficulty: many
571 routes are still available. Indeed, a key finding of this research is that fitness
572 disadvantages of intermediate mutants sometimes have a great effect on the
573 proportion of adapted mutant, but only when a large number of intermediate
574 mutants are deleterious, and their fitness cost is large.

575

576 Although some studies recognize the importance of deleterious intermediates
577 and the crossing of fitness valleys to the overall adaptive evolution of pathogens
578 (27, 29, 42, 43), and the possibility of multiple simultaneous mutations to
579 overcome such fitness valleys (44), various other models assume a strong-
580 selection-weak-mutation paradigm (32–34, 45), ignoring any adaptive trajectories
581 that require the crossing of a fitness valley. Such assumptions might be
582 appropriate for small population sizes, or pathogens with low mutation rates (5,
583 30). However, for a pathogen with a large population size and high mutation rate,
584 these SSWM assumptions are substantially violated. For influenza virus for
585 example, the mutation rate is around 1 mutation per 10^5 nucleotides, per round of
586 genome replication ($r = 10^{-5}$) (37, 38), and the population size P easily exceeds

587 10^{10} virions in a single host, hence the SSWM conditions $4rP \ll 1$ or
588 $rP \ll 1/\ln(Ps)$, where s is the fitness increase, are not fulfilled (30, 46).

589

590 It should be noted though, that the assumptions of the deterministic
591 approximation are violated in parts of *Tables 1* and *2* (which showed the
592 numbers of routes for scenarios where up to 7 mutations are required). When the
593 inverse of the error rate to the power of the number of mutations required ($1/r^m$) is
594 larger than or comparable to the population size, stochastic variations may
595 become relevant. This was seen in *Figure 1A*, where for a population size around
596 10^{10} the two mutations were not acquired as readily as the deterministic model
597 would have suggested. The smaller the population size is in comparison to $(1/r^m)$,
598 the higher the likelihood that stochastic effects decrease the expected proportion
599 observed in the pathogen population. Especially when large numbers of
600 mutations are required, the expected number of times the adapted mutant occurs
601 will be small, if not zero, when taking account of the population size. Moreover,
602 stochastic death of rare intermediate or fully adapted mutants will further affect
603 the observed proportions. In a single host, one can multiply the effective virus
604 population size, say 10^{10} , with the probability of interest, e.g. 6.65×10^{-7} for a
605 random virion being a double mutant when the intermediate single mutants were
606 fully deleterious, the starting mutant neutral and the double mutant has a fitness f
607 $= 1.5$, to get the expected number of virions with the mutations of interest, here
608 6650. In the context of transmission, where frequently small populations
609 consisting of less than a handful of virions are estimated to start new infections

610 for e.g. influenza virus, HIV and hepatitis C (47–50), the probability of any virion
611 being a mutant of interest is informative for epidemiological studies and risk
612 assessment. In the biologically implausible case (but just to clarify) that each
613 virion has the same chance of being transmitted and starting the next infection, if
614 only a single virion starts the next infection, the chance that the next host is
615 infected with only the adapted virus is identical to the proportion of this adapted
616 virus in the donor host.

617

618 Interestingly, evolutionary models have also been used to improve the
619 understanding of the developmental stages and processes in cancer, and to
620 increase the efficacy of treatment regimes (51). In the evolution of a cancerous
621 cell, there is often a fitness valley to be crossed before the cell is able to progress
622 to expansive, uncontrolled growth (51). As with many other evolutionary models,
623 models for cancer evolution are focused on population-level adaptation. The
624 cancer literature describes two main mechanisms for the population-level
625 crossing of the fitness valley: sequential fixation, whereby the full cancerous cell
626 population acquires one mutation, and only after fixation of the first mutation, the
627 second mutation becomes fixated; and stochastic tunneling, whereby the second
628 mutation establishes prior to fixation of the first mutation (52, 53). Stochastic
629 tunneling describes the probability of fixation on a population-level, and allows for
630 sequential but not necessarily simultaneous mutations, in contrast, the
631 deterministic calculations above describe the probability of obtaining multiple
632 mutations simultaneously by any single unit (cell or pathogen or virion), and can

633 be converted to an expected proportion in the population, but does not equate to
634 fixation.

635

636 An advantage of the deterministic calculations used here is that this approach
637 can easily be adjusted to encompass more complicated schemes of required
638 mutations and associated fitnesses. It is, for example, not limited to the
639 investigation of effects of deleterious intermediates, but can also be used when
640 individual mutations are neutral or beneficial, and the combined mutation is
641 synergistic, for example mutations at positions 138 and 229 in the non-structural
642 protein 1 (NS1) (54) and 147, 339 and 588 in PB2 (55) of influenza A/H5N1 virus
643 affecting virulence.

644

645 The implementation can be easily changed to model other situations, for example
646 i) where a mutation has a fitness effect when in the vRNA, but not when
647 occurring in the cRNA, as those molecules are not transcribed into mRNA and
648 translated into protein; or ii) to encompass delayed phenotypes of mutations (56,
649 57) whereby deleterious or advantageous fitness effects are not fully observed as
650 the respective proteins are only generated in a meaningful amount at a later time.
651 Such mechanisms might alter the likelihood of deleterious and adaptive
652 mutations occurring, for example by a deleterious mutation arising as non-
653 deleterious in the cRNA, and the compensatory mutation arising in the next
654 replication round, such that deleterious vRNA is never formed and both
655 mutations were effectively neutral.

656

657 In the stochastic model, the branching factor governs population growth, and as
658 a result of the founder effect this leads to mutations arising earlier in time
659 achieving higher proportions; an effect that can be seen as a banded simulation
660 runs, for example in Figure 4A, where the top band shows that there are fewer
661 instances (lower red intensity) but higher proportions of double mutants for
662 mutants arising early. Again, the implementation of this model can be adjusted
663 such that the branching factor varies in both steps of replication to match the
664 specific parameters for the virus of interest.

665

666 The calculations enable both estimating the likelihood of crossing fitness valleys,
667 as well as the probability of passing a narrow fitness ridge. The work presented
668 here on assessing the effect of fitness ridges and required order is relevant, for
669 example, in modeling antibiotic resistance (26), and pyrimethamine resistance of
670 malaria (58). The equations can easily incorporate variable mutation rates (59),
671 which may be useful to investigate different polymerase error rates of influenza
672 virus (60), and to account for varying replication fidelity of HIV reverse
673 transcriptase along different positions in the genome (61). This feature is also
674 important in the evolutionary modeling of cancer, where disease progression
675 often involves the acquisition of decreased genetic stability, and thus an altered
676 mutation rate (62).

677

678 Counting the number of routes is also a method that could be applied to
679 determine the multiplicity of drug therapy, as the acquisition of drug-resistant
680 mutations might be avoided by a treatment regime shaping the fitness valley
681 deep and wide enough to prevent the pathogen from crossing, an approach that
682 has been described with combination therapy for example in influenza,
683 tuberculosis and HIV treatment (61, 63–65). In the context of drug therapy,
684 Ribeiro *et al.* already noted that for totally defective intermediate HIV mutants, all
685 higher order strains have to be produced directly from wild-type, i.e. only allowing
686 routes where all mutations are acquired at once; they also described that for
687 smaller selective disadvantages, a $k+1$ mutant is most likely produced from a k -
688 point mutant, i.e. a qualitative phrasing of our quantitation of the contribution to
689 the number of routes from obtaining two subsequent mutations for less-
690 deleterious intermediates (66).

691

692 The introduction of fitness valleys can also be exploited as a mitigation strategy
693 for infectious diseases. One could, for example, design vaccines that require a
694 pathogen to obtain destabilizing mutations to enable immune escape. Models of
695 pathogen evolution can help to establish whether such approaches will
696 completely stall adaptation of the pathogen, or with what likelihood the created
697 fitness valleys would be crossed. Moreover, such approaches could also be used
698 to explore alternative routes as a result of epistatic interactions that might allow
699 deleterious mutations to occur if acquired in the right order (15, 17, 20).

700

701 The successful and efficient invasion of zoonoses into the human population is
702 often thought to be constrained by the existence of deleterious mutations on the
703 path to adaptation. Therefore, calculations on the effects of fitness valleys are of
704 critical importance in pandemic risk assessment of emerging pathogens (8, 33,
705 35, 44, 67, 68), and in addition to inform the cost-benefit analyses of gain-of-
706 function experiments and dual-use research of concern.

707

708 In summary, the ability to calculate the effect of deleterious mutations and order,
709 and to understand the results with the description of the number of available
710 routes, helps to assess the expected impact of fitness valleys and ridges in
711 pathogen evolution, with applications in drug resistance, immune escape, and
712 zoonotic risks assessments.

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- 929

930 **FIGURES:**

931 **Figure 1: A)** The blue line shows the deterministic probability of any virion being
932 a double mutant (\log_{10}) as a function of the relative fitness of the intermediate
933 single mutants (a relative fitness of 0 means the single mutants are fully
934 deleterious). The probability is shown after $t = 20$ replication rounds, in the
935 situation where the starting (wild type) virus and the adapted (double) mutant are
936 neutral (relative fitness = 1). The pale shades of blue indicate the results of
937 10,000 stochastic simulations for each of the 101 settings of the relative fitness f .
938 The circles indicate the average of the stochastic runs for each fitness setting. **B)**
939 The probability of any virion being a double mutant is split into a mechanism
940 where two mutations were acquired in a single replication round ("through
941 doubles"); and a mechanism where the two single mutations occurred in distinct
942 replication rounds ("through singles").

943

944 **Figure 2:** The effective number of routes to a double mutant is shown as a
945 function of the relative fitness of the intermediate single mutants (a relative
946 fitness $f = 0$ means the single mutants are fully deleterious). The results are
947 shown after $t = 20$ replication rounds, in the situation where the starting (wild
948 type) virus and the adapted (double) mutant are neutral ($f = 1$). The colors
949 illustrate the relative contributions of a double mutation at once (grey, delay = 0),
950 and the single-single mutation routes, with an increasing delay of 1 until 19

951 replication rounds between the mutation events represented by the gradient
952 shown in the colorbar.

953

954 **Figure 3:** The probability of any virion being a double mutant (\log_{10}) is shown in
955 color as a function of the relative fitness of the deleterious intermediate single
956 mutants (x-axis) and the relative fitness of the double, i.e. host-adapted, mutant
957 (y-axis). The probability is shown after $t = 20$ replication rounds, and the starting
958 (wild type) virus is neutral ($f = 1$).

959

960 **Figure 4: A)** The deterministic probability of any virion being a double mutant
961 (\log_{10}) is shown as a function of the relative fitness of the deleterious
962 intermediate single mutant(s). The graph shows the situation when both single
963 mutants are deleterious (“fitness valley”, see the blue line in *Figure 1*), or when
964 only the non-ordered single mutant is deleterious (“fitness ridge”). The probability
965 is shown after $t = 20$ replication rounds, the starting (wild type) virus, ordered
966 single mutant (for the fitness ridge) and double mutant are neutral ($f = 1$). The
967 pale shades of blue and red indicate the results of 10,000 stochastic simulations
968 for each of the 101 settings of the relative fitness f for the fitness valley, and
969 fitness ridge, respectively. The circles indicate the average of the stochastic runs
970 for each fitness setting. **B)** The probability of double mutant for the fitness ridge
971 (red line) is divided into the contribution toward this probability by the mechanism
972 where two mutations were acquired in a single replication round (“through
973 doubles”); as two single mutations in distinct replication rounds in order (“through

974 ordered singles"); and as two single mutations in distinct replication rounds
975 occurring in the incorrect order, where the deleterious single mutation is obtained
976 first and incurs the fitness cost ("through deleterious singles").

977 **TABLES**

978

	# mutations required					
	2	3	4	5	6	7
Effective # of routes	20	20	20	20	20	20
Total routes	400	8×10^3	1.6×10^5	3.2×10^6	6.4×10^7	1.28×10^9
Fraction available	5.0×10^{-2}	2.5×10^{-3}	1.3×10^{-4}	6.3×10^{-6}	3.1×10^{-7}	1.6×10^{-8}

979

980 **Table 1:** The effective number of routes if all intermediate mutants are fully
 981 deleterious, the number of available routes if all intermediate mutants were viable
 982 (total routes), and the fraction of routes available (Effective # of routes/Total
 983 routes). The results are shown for two to seven mutations required, for $t = 20$
 984 replication rounds. The starting (wild type) virus and the adapted (double) mutant
 985 are neutral ($f = 1$), and all possible intermediates are fully deleterious ($f = 0$).

986

987

	# mutations required					
	2	3	4	5	6	7
Allowed routes	210	1540	8855	42504	177100	657800
Total routes	400	8×10^3	1.6×10^5	3.2×10^6	6.4×10^7	1.28×10^9
Fraction available	0.525	0.193	0.055	0.013	0.003	5.14×10^{-4}

988

989 **Table 2:** The number of allowed routes if complete and strict order is required for
990 all mutations, the number of available routes if all intermediate mutants were
991 viable (total routes), and the fraction of routes available, for situations where two
992 to seven mutations are required. As an example, for 4 mutations required, this
993 would mean that mutation A has to occur prior to or simultaneously with mutation
994 B; mutation B has to occur prior to or simultaneously with mutation C; and
995 mutation C has to occur prior to or simultaneously with mutation D. The results
996 are shown for $t = 20$ replication rounds, the starting (wild type) virus and each
997 ordered mutant are neutral ($f = 1$), and all possible non-ordered intermediate
998 mutants are fully deleterious ($f = 0$).

999