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Phylogenetic interpretation during outbreaks requires caution

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How viruses are related, and how they have evolved and spread over time, can be investigated using phylogenetics. Here, we set out how genomic analyses should be used during an epidemic and propose that phylogenetic insights from the early stages of an outbreak should heed all the available epidemiological information.

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A goal of genomic epidemiology is to infer epidemiological and emergence dynamics from virus genome sequences obtained over short epidemic timescales ¹. Rapid in situ sequence generation and phylogenetic inference is based on detection of genetic changes in pathogen sequences. But during outbreaks there are many unknowns. The outbreak of coronavirus disease 2019 (COVID-19), which originated in Wuhan, China, was reported in December 2019 ². By January 2020, the genome of the causative novel coronavirus, named SARS-CoV-2, had been sequenced and made publicly available ². Virus sequences have

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26 underpinned development of diagnostics and vaccines and been used to assess patterns of
27 transmission and spread. Although sequence data was used to answer crucial epidemiological
28 questions during the Ebola and Zika outbreaks ^{3,4}, the pace of generation of SARS CoV-2
29 genome data generation is unprecedented and is informing public health policy in real-time.

30 Importantly, it's not only sequences that inform phylogenies, and multiple factors
31 contribute to the outputs including model assumptions, sampling density, timing of sample
32 collection, portion of the viral genome sequenced, quality of sequencing data and the
33 mutation rate of the virus itself. Although it is important to extract as much information as
34 possible from sequence data as outbreaks unfold, it is imperative to bear in mind that the
35 historical relationships of strains (phylogenies) are hypotheses that can be challenged as more
36 data becomes available. Here, we highlight some of the challenges of genomic epidemiology
37 during outbreaks such as SARS-CoV-2 and advise that interpretation of findings from
38 phylogenies needs to assess all epidemiological and supporting information and consider
39 sources of bias.

40 During outbreaks we want to know if cases are linked and if this implies transmission.
41 Most viruses can be separated into strains and if two infections are caused by dissimilar
42 strains one can rule out transmission. The oft-forgotten point is that phylogenies can rule out
43 transmission, but if infections are caused by the same strains or identical viruses it does not
44 decisively prove transmission. During an emerging outbreak, when pathogens have not yet
45 diverged into different strains, phylogenetic information is too weak to hypothesize
46 transmission linkage—which in turn can be used for geographic inference; even if the
47 phylogenetic information is stronger, the same phylogeny is consistent with multiple
48 transmission histories and there may missing links due to incomplete sampling ⁵.
49 Consequently, we need to combine phylogenetic findings with epidemiological and
50 supporting information such as environmental factors and human air travel data before we
51 draw any immediate conclusions regarding transmission. This was the case with Zika virus in
52 Africa where epidemiological, human mobility and climatic data supported the phylogenetic
53 hypothesis that the outbreak was likely imported from Brazil ⁶.

54 In the first stage of an outbreak, we can use phylogenetics to discern possible zoonotic
55 sources, as in the case of the 2018 Lassa fever virus outbreak, where phylogenetic patterns
56 indicated independent spillover events from rodent hosts ⁷. The crucial observation was that
57 the correct identification of the source of zoonotic transmission relies on the availability of

58 viral genome sequences from potential animal reservoirs. If the source of any virus has not
59 been sampled, it cannot be inferred, because phylogenetic linkage alone does not prove it.
60 This is the reason for uncertainty surrounding the zoonotic source of SARS-CoV-2, because
61 we have limited knowledge about the viral abundance from potential animal reservoirs ⁸. The
62 generation of additional viral genome sequences from an outbreak, coupled with virus-
63 specific and epidemiological knowledge, provides insight into whether or not multiple
64 ‘jumps’ occurred from a reservoir that might warrant appropriate control measures. Identical
65 or nearly-identical virus genomes are expected from early transmission chains if a single
66 spillover occurred recently, unless multiple zoonoses originated from the same low-genetic
67 diversity virus pool. In contrast, higher diversity in the early-stage of human-to-human
68 transmission is expected if multiple zoonoses have occurred or if there is significant within-
69 host evolution ⁹.

70 Geographical inferences (where and when) are feasible as more representative viral
71 genome data—in temporal and spatial scales—becomes available. We can hypothesize the
72 location of common ancestors using ancestral reconstruction methods and infer phylogenies
73 scaled to time, in order to date epidemiological events. Such analyses require a molecular
74 clock, which models how the rate with which mutations accumulate with time, and how this
75 varies across the branches of a phylogeny. However, early in an outbreak there may not be
76 sufficient signal to accurately estimate clock rate. If this is the case, then it might be
77 appropriate to apply an estimate from another closely related virus ¹⁰. If temporal signal is
78 present and a clock rate can be estimated, results need to be reported as credible intervals
79 (instead of point estimates) to account for uncertainty in both the data (incomplete, biased, or
80 improper sampling can lead to misleading phylogenies) and the many aspects of the methods.

81 When investigating the dissemination of an emerging virus the number of sequenced
82 viral genomes may not be representative. Even as the outbreak unfolds, and more genomes
83 are obtained, they only represent a snapshot of the underlying genetic diversity. If
84 phylogenies are considered alone we cannot conclusively assert the geographical origins of
85 the virus—or the extent of community transmission—as we cannot distinguish between local
86 transmission events and multiple introductions of genetically similar viruses, from
87 geographically distinct sources, if one of them has not been sampled. In this way uneven
88 sampling can also lead to misleading conclusions on the geographical source, number of
89 introductions and the size and duration of local transmission chains ¹¹. The significance of
90 these associations is harder to ascertain when the phylogeny is reported without any

91 assessment on the reliability of internal branches. Therefore, phylogenetic interpretation from
92 ongoing outbreaks as is the case of SARS-CoV-2 needs to be done in the context of all
93 available information such as temporal and spatial distribution of cases, travel patterns and
94 any evidence of epidemiological linkage, sampling uncertainty and other sources of bias need
95 to be carefully considered and reported.

96 The methods for valid phylogenetic inference require multiple assumptions which are
97 likely not met during emerging outbreaks. Examples (not exhaustive) include adequate
98 phylogenetic signal, which is low when strains have not yet diverged; geographical
99 representation and effective sampling time points with sufficient molecular clock signal,
100 which only become feasible as the epidemic unfolds; and random mixing, which may be
101 violated under certain circumstances, for instance when mitigation strategies are set in place.
102 Estimates from phylogenies may be sensitive to one or more of these assumptions and
103 conclusions need to be made and shared with caution. Another essential consideration during
104 an epidemic is accurate rooting of the phylogeny as it determines the direction of
105 transmission over time ¹².

106 There are also genome features that are intrinsic to the biology of the virus that may
107 impact the extent and applicability of phylogenetics during outbreaks. For instance, the
108 presence of recombination/reassortment and low diversity (due to the rate of evolution,
109 selective constraints and transmission bottlenecks) complicate the resolution of phylogenetic
110 relationships, but the incorporation of within-host viral diversity may provide greater
111 resolution in understanding transmission dynamics ¹³. Moreover, some of mutations in the
112 viral genome sequence can be due to the error rate of the sequencing technology, recurrent
113 sequencing issues, hypermutability or contamination which warrant caution with
114 interpretations and especially with those concerning selection and recombination.

115 Genomic epidemiology has supported public health outbreak responses. Indeed, the
116 ability to exploit viral genome sequences has allowed us to characterise early patterns of
117 SARS-CoV-2 transmission in China, New Zealand and Australia ^{14,15}. In the midst of an
118 outbreak sharing data is both necessary and important for an effective response, but sharing
119 the associated metadata is also necessary to aid interpretations (e.g. how representative is the
120 data of the country-wide situation) and to avoid creating sampling bias by researchers that are
121 not doing the sequencing themselves.

122 The emergence of SARS-CoV-2 has presented a series of challenges about how we
123 reliably extract information from phylogenies to gain insights into virus transmission and
124 spread, and how we responsibly present our findings. Owing to low genetic diversity and
125 uneven sampling, several controversial hypotheses have already been put forward. One
126 cautionary tale involves how an outbreak in Bavaria seeded the epidemic in northern Italy
127 and the subsequent wider outbreak in Europe. This notion was based on a small sample of
128 very similar sequences. However, it overlooked a more likely scenario in which this virus
129 was already circulating in China and that European regions had multiple introductions from
130 China. At this early stage conclusions about the impacts of mutations on transmission and
131 disease (e.g. D614G mutation in the spike protein ¹⁶) should not be made on the basis of
132 phylogenies alone but with separate evidence supporting not only a phenotypic difference but
133 the resulting consequences for epidemiology.

134 The SARS-CoV-2 pandemic has highlighted the importance of providing a
135 comprehensive rationale for any conclusions about the spatio-temporal dispersal of the virus.
136 Phylogenies represent hypotheses that encompass different sources of error and this
137 uncertainty needs to be visualised and communicated far more transparently. Another
138 challenge is how we facilitate the dissemination of metadata and integrate this with
139 phylogenetic trees. Incorporating host characteristics (e.g. age, onset date, exposure history)
140 to aid phylogenetic interpretation would undoubtedly results in more reliable inferences.

141 Now, more than ever, careful reporting of phylogenetic interpretations, while
142 safeguarding the privacy of infected individuals, would ensure that both policymakers and the
143 public have the best possible information during an outbreak. Failure to balance these issues
144 could jeopardise both scientific integrity and public confidence in the field of genomic
145 epidemiology.

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147 **REFERENCES**

- 148 1. Grubaugh, N. D. et al. *Nat Microbiol* **4**, 10–19 (2019).
- 149 2. Wu, F. et al. *Nature* **579**, 265–269 (2020).
- 150 3. Holmes, E. C., Dudas, G., Rambaut, A. & Andersen, K. G. *Nature* **538**, 193–200

- 151 (2016).
- 152 4. Pollett, S. et al. *J. Infect. Dis.* (2019).
- 153 5. Hall, M. D. & Colijn, C. *Mol. Biol. Evol.* **36**, 1333–1343 (2019).
- 154 6. Hill, S. C. et al. *Lancet Infect. Dis.* **19**, 1138–1147 (2019).
- 155 7. Kafetzopoulou, L. E. et al. *Science* (80). **363**, 74–77 (2019).
- 156 8. Andersen, K. G., Rambaut, A., Ian Lipkin, W., Holmes, E. C. & Garry, R. F. *Nat.*
157 *Med.* 1–3 (2020).
- 158 9. Didelot, X., Fraser, C., Gardy, J. & Colijn, C. *Mol. Biol. Evol.* **34**, 997–1007 (2017).
- 159 10. Fraser, C. et al. *Science* (80). **324**, 1557–1561 (2009).
- 160 11. Kraemer, M. U. G. et al. *Epidemiol. Infect.* **147**, (2019).
- 161 12. Dudas, G. & Rambaut, A. *PLoS Curr.* **6**, (2014).
- 162 13. Worby, C. J., Lipsitch, M. & Hanage, W. P. *Am. J. Epidemiol.* **186**, 1209–1216
163 (2017).
- 164 14. Lu, J. et al. *Cell* (2020). doi:10.1016/j.cell.2020.04.023
- 165 15. Eden, J.-S. et al. *Virus Evol* **6**, veaa027 (2020).
- 166 16. Korber, B. et al. bioRxiv 2020.04.29.069054 (2020). doi:10.1101/2020.04.29.069054

167 **Contributions**

168 D.C.T. conceived the commentary and wrote the first draft. C.J-V.A, D.C.T conceptualized
169 the ideas with W.P.H. All authors edited the manuscript into its final form.

170 **Competing Interests**

171 The authors declare no competing interests.