Spatial Growth Rate of Emerging SARS-CoV-2 Lineages in England, September 2020–December 2021

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Abstract

This paper uses a robust method of spatial epidemiological analysis to assess the spatial growth rate of multiple lineages of SARS-CoV-2 in the local authority areas of England, September 2020–December 2021. Using the genomic surveillance records of the COVID-19 Genomics UK (COG-UK) Consortium, the analysis identifies a substantial (7.6-fold) difference in the average rate of spatial growth of 37 sample lineages, from the slowest (Delta AY.4.3) to the fastest (Omicron BA.1). Spatial growth of the Omicron (B.1.1.529 and BA) variant was found to be 2.81× faster than the Delta (B.1.617.2 and AY) variant and 3.76× faster than the Alpha (B.1.1.7 and Q) variant. In addition to AY.4.2 (a designated variant under investigation, VUI-21OCT-01), three Delta sublineages (AY.43, AY.98 and AY.120) were found to display a statistically faster rate of spatial growth than the parent lineage and would seem to merit further investigation. We suggest that the monitoring of spatial growth rates is a potentially valuable adjunct to routine assessments of the growth of emerging SARS-CoV-2 lineages in a defined population.

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Introduction

Emerging lineages of the severe acute respiratory syndrome coronavirus 2 (SARS-CoV-2) have the potential to place significant pressure on public health systems due to increased infectivity, transmissibility, virulence, immune escape or other fitness advantage [1, 2]. Global genomic surveillance has identified >1,700 SARS-CoV-2 lineages since the beginning of the COVID-19 pandemic [3, 4], of which Alpha (B.1.1.7 and Q), Beta (B.1.351), Gamma (P.1), Delta (B.1.617.2 and AY) and Omicron (B.1.1.529 and BA) have been designated as variants of concern by the World Health Organization (WHO) on account of their global public health significance [5]. Additional lineages are currently classified on the basis of properties that are suggestive of an emerging (variants of interest) or possible future (variants under monitoring) risk to global public health [5]. The risk is well illustrated by the recent and rapid emergence of Omicron as the dominant variant in the United Kingdom, South Africa and the United States, among other countries, in late November and December 2021 [6–8].

One important epidemiological facet of an emerging SARS-CoV-2 lineage is its propensity to grow in a defined population [9]. There are well-established methods for assessing the rate of temporal growth by, for example, examining the trajectory of case doubling times or estimating the basic reproduction number, R_0 , of the agent in question [10, 11]. Viewed from a geographical perspective, these measures are essentially aspatial in that they provide very little information on the geographical growth, or spatial expansion, of the associated infection wave. To extend the examination of SARS-CoV-2 growth rates into the spatial domain, the present paper applies a robust method of spatial epidemiological analysis that is known as the *swash-backwash model of the single epidemic wave* [12] to the genomic surveillance records of the COVID-19 Genomics UK (COG-UK) Consortium [13]. Using the spatial sequence of detection of sample variants as a proxy for the spatial wave front of infection, our examination yields estimates of the spatial growth rate of multiple SARS-CoV-2 lineages in the local authority areas of England, September 2020—December 2021.

For a total of 37 sample lineages under investigation, we present evidence of a substantial (7.6-fold) difference in the average rate of spatial growth, from the slowest (Delta

AY.4.3) to the fastest (Omicron BA.1). Whilst the overall results for the Alpha, Delta and Omicron variants are consistent with the documented growth advantages for these lineages, several emergent Delta sublineages (AY.4.2, AY.43, AY.98 and AY.120) are found to have had a statistically significant growth advantage over the parent lineage. To our knowledge, this is the first comparative study of the spatial growth rate of multiple emerging SARS-CoV-2 lineages at the national level. It is also the first report of a spatial growth advantage for the Delta AY.43, AY.98 and AY.120 lineages, and the first to document an apparently reduced spatial growth rate for a substantial number of other AY lineages that emerged in the spring and summer of 2021. We suggest that the monitoring of spatial growth rates is a potentially valuable adjunct to routine assessments of the growth of SARS-CoV-2 lineages in a defined population.

Data and methods

Since September 2020, successive waves of SARS-CoV-2 infection with emerging lineages of the Alpha (September 2020 onset), Delta (March 2021 onset) and Omicron (November 2021 onset) variants have been recorded in England [14–16]. The weekly record of COVID-19 cases to mid-December 2021 is plotted in Figure 1A, whilst the underpinning sequence of variants is depicted in Figure 1B. As Figure 1B shows, Alpha, Delta and Omicron achieved the status of dominant variants in December 2020, May 2021 and December 2021, respectively.

FIGURE 1 NEAR HERE

Data

We draw on the integrated national-level SARS-CoV-2 genomic surveillance records of the COVID-19 Genomics UK (COG-UK) Consortium [13]. Lineage counts by local authority area and week of sample collection for England were accessed from the COG-UK website [18] for a 68-week period, September 2020 (epidemiological week 36, ending 5 September) to December 2021 (epidemiological week 50, ending 18 December) (Figure 1B). The data set included geo-coded information on 979,075 SARS-CoV-2 samples assigned to the 309

Lower Tier Local Authority (LTLA) divisions of England. Here, we define the 309 LTLAs according to their most recent (May 2021) status. Information on the lineage of 20,655 samples (2.1%) was either suppressed (1,105) or not recorded (19,550). Of the remaining 958,420 samples, the majority (93.8%) were classified as belonging to the B.1.1.7 and Q (Alpha, 153,405 samples), B.1.617.2 and AY (Delta, 722,133 samples) and B.1.1.529 and BA (Omicron, 23,137 samples) lineages (Table 1). Samples belonging to these lineages form the basis of all our analysis.

TABLE 1 NEAR HERE

Methods

To assess the spatial growth rate of a given SARS-CoV-2 lineage, we draw on the *swash-backwash model of the single epidemic wave* [12]. In essence, the model represents a spatial derivative of the generic *SIR* mass action models of infectious disease transmission [19]. Using the binary (presence/absence) of a disease, the model (i) allows the disaggregation of an infection wave into phases of spatial expansion and retreat and (ii) provides a means of measuring the phase transitions of geographical units from susceptible *S*, through infective *I* to recovered *R* status. See, for example, Smallman-Raynor and Cliff [20] and Smallman-Raynor, *et al.* [21].

Measuring the spatial growth rate

Full details of the modelling procedure are outlined by Cliff and Haggett [12]. For the purposes of the present analysis, we focus on the spatial expansion phase (that is, the change of state from S to I across a set geographical units) for a given SARS-CoV-2 lineage. Specifically, let the first week in which the lineage was detected in England be coded as t = 1. Subsequent weeks were then coded serially as t = 2, 3, ..., T, where T is the number of weekly periods from the beginning to the end of the detected occurrence of the lineage. For any given geographical unit, we refer to the first week in which the lineage was detected as the *leading edge* (LE) of the infection wave. The average time (in weeks) to the detection of

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the lineage across the set of units can then be defined by a time-weighted mean, \bar{t}_{LE} , of the form

$$\bar{t}_{LE} = \frac{1}{N} \sum_{t=1}^{T} t n_t.$$
 (1)

Here, n_t is the number of units whose leading edge, LE, occurred in week t and $N = \sum n_t$. Formed in this manner, SARS-CoV-2 lineages with relatively high rates of spatial expansion (or rapidly developing LE) take on relatively low values of \bar{t}_{LE} (i.e. short average times to detection). Conversely, lineages with relatively low rates of spatial expansion (or slowly developing LE) take on relatively high values of \bar{t}_{LE} (i.e. long average times to detection).

Application of the model

Equation (1) was used to estimate the spatial growth rate of sample SARS-CoV-2 lineages for which the earliest detection in England occurred in the time period covered by the dataset (September 2020–December 2021) and for which substantial geographical spread had been documented. To ensure the inclusion of sufficient observations for geographical analysis, the sample was limited to lineages that had been detected in at least one-third of the 309 LTLAs by December 2021. Based on these criteria, the sample consisted of 37 lineages. Summary details of the sample, including the number of LTLAs in which each lineage was detected, the total count of detections over the study period, and the earliest date of detection, are provided in Table 1.

For each lineage, equation (1) was fitted with t = 1 set to the week of earliest detection in Table 1. In the instance of Omicron, retrospective analysis has identified a lone detection of the BA.1 lineage in epidemiological week 43 of 2021 (week ending 30 October), four weeks prior to the subsequent detection and apparent onset of widespread transmission of the variant in epidemiological week 47 (week ending 27 November). For the purposes of the present analysis, we set week 47 as t = 1 for Omicron, but we also report the computed value of \bar{t}_{LE} based on the earlier detection in week 43. Finally, we exclude two LTLAs (City of London and Isles of Scilly) from all analysis on account of the suppression of lineage data due to their small populations. Data analysis was performed in Minitab[®]17 (Minitab Inc.,

Pennsylvania) and data mapping in QGIS 3.10.14-A Coruña (QGIS.org) using Local Authority Districts (May 2021) UK and Regions (December 2020) EN shapefiles from the Office for National Statistics (ONS) [22].

Results

Table 1 confirms that the 37 sample lineages were geographically extensive in their transmission, with 29 having been detected in >150 LTLAs, 21 in >250 LTLAs, 16 in >300 LTLAs, and nine in the complete set of 307 LTLAs under examination. The majority (23) were associated with >1,000 detections, 13 with >10,000 detections and three with >100,000 detections. Delta (B.1.617.2 and AY) was the most common lineage (722,133 detections) and AY.4 the most common sublineage (547,403 detections), with AY lineages accounting for 33 of the spread events under examination. In turn, the majority of lineages emerged (as judged by the date of earliest detections) in the spring and summer of 2021, as the Delta infection wave was evolving both domestically and internationally.

Spatial growth curves and leading edge (*LE*) maps

The upper graphs in Figure 2 plot the count of LTLAs by week of earliest detection of the Alpha (B.1.1.7 and Q), Delta (B.1.617.2 and AY) and Omicron (B.1.1.529 and BA) variants, where weeks are indexed to the earliest detection of the respective variants (Table 1). The lower graphs are spatial growth curves, formed by replotting the information in the upper graphs as a cumulative proportion of LTLAs. Average curves for the set of sample lineages in Table 1 are shown for reference.

FIGURE 2 NEAR HERE

Together, the graphs in Figure 2 portray the temporal development of the spatial leading edges (LE) for each variant. The geographical expression of these LE is captured by the choropleth maps in Figure 3 which plot the week of earliest detection of each variant in the set of LTLAs. The sequentially more rapid spatial growth of the variants (Alpha \rightarrow Delta \rightarrow Omicron) is evidenced by the sequentially steeper spatial growth curves (Figure 2) and the

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sequentially shorter periods to earliest detection (Figure 3). The latter feature is emphasized when earliest detections are formed as regional averages in Figure 4.

FIGURE 3 NEAR HERE

FIGURE 4 NEAR HERE

Rates of spatial growth (\bar{t}_{LE})

The right-hand column in Table 1 summarizes the results of the application of equation (1) to each of the sample lineages. Computed values of \bar{t}_{LE} and associated 95% confidence intervals (95% CI) are given, along with an overall average value of \bar{t}_{LE} for the entire sample. As noted above, lineages with relatively *high* rates of spatial expansion (or rapidly developing *LE*) are represented by relatively *low* values of \bar{t}_{LE} (i.e. short average times to detection), while lineages with relatively *low* rates of spatial expansion (or slowly developing *LE*) take on relatively *high* values of \bar{t}_{LE} (i.e. long average times to detection). In this manner, the table confirms the sequential increase in the spatial growth rate for Alpha, Delta and Omicron. On average, the earliest detection of the Alpha variant in a given LTLA occurred at $\bar{t}_{LE} = 9.90$ (95% CI: 9.56, 10.23) weeks after the earliest sampled detection in England. This reduced to 7.40 (95% CI: 7.12, 7.68) weeks for Delta and 2.63 (95% CI: 2.56, 2.71) weeks for Omicron.

Delta AY lineages

Figure 5 is based on the information in Table 1 and plots the values of \bar{t}_{LE} for B.1.617.2 and AY lineages in order, from the lowest (left, high values of \bar{t}_{LE}) to the highest (right, low values of \bar{t}_{LE}) rates of spatial growth. Values are plotted on an inverted vertical scale to facilitate interpretation. The average value of \bar{t}_{LE} , formed across the sample set of lineages in Table 1, is indicated for reference as are the \bar{t}_{LE} for the Alpha (B.1.1.7 and Q), Delta (B.1.617.2 and AY) and Omicron (B.1.1.529 and BA) variants. Spatial growth curves, formed in the manner of Figure 2, are plotted for a sample of 20 AY lineages with relatively high and low rates of spatial growth in Figure 6.

FIGURE 5 NEAR HERE

FIGURE 6 NEAR HERE

There is a 7.6-fold difference in the range of values of \bar{t}_{LE} in Figure 5, from Delta AY.4.3 with the lowest spatial growth rate (19.93 weeks) to Omicron with the highest (2.63 weeks). A group of four AY lineages (AY.4.2, AY.43, AY.98 and AY.120), first detected in the period from mid-May to mid-July 2021, are positioned between Delta and Omicron in Figure 5 and display rates of spatial growth that are significantly higher (as judged by 95% CI) than the aggregate rate for the Delta variant. In contrast, the overwhelming majority of AY lineages display statistically lower – in many instances substantially lower – spatial growth rates (as judged by 95% CI) than the aggregate rate for the Delta variant.

Discussion

Recent experience has underscored the importance of the ongoing tracking, monitoring and analysis of emerging SARS-CoV-2 lineages with a view to mitigating the impacts of the COVID-19 pandemic [23]. We have used a robust model of spatial epidemiological analysis to estimate the rate of spatial growth of multiple lineages of the virus in England over a 68-week period, September 2020–December 2021. We have shown that the Alpha, Delta and Omicron variants took an average of 9.90 weeks, 7.40 weeks and 2.63 weeks, respectively, to reach the set of LTLAs under examination (Table 1 and Figure 5). Expressed in relative terms, the leading spatial edges were 1.34× faster (Delta vs. Alpha), 2.81× faster (Omicron vs. Delta), and 3.76× faster (Omicron vs. Alpha). Our estimates scale to the approximate length of time that Alpha (12 weeks), Delta (8 weeks) and Omicron (3 weeks) took to establish themselves as the dominant variants in England [18], and are consistent with evidence for the fitness advantage of Delta over Alpha and Omicron over Delta [11, 24, 25].

Of the 121 Delta AY lineages detected in England to December 2021 and included in the genomic surveillance records of the COG-UK Consortium, 33 lineages met the geographical criterion for inclusion in the current analysis. In interpreting the results for these lineages, we note that AY designations are phylogenetically defined and do not necessarily denote any fundamental biological differences between the lineages [26]. Moreover, results of the type documented in this paper are context dependent and cannot be interpreted as evidence of a change in biological transmissibility, immune escape or other fitness advantage.

Subject to these caveats, we have identified four AY lineages (AY.4.2, AY.43, AY.98 and AY.120) for which the rate of spatial growth exceeded the aggregate rate for the Delta variant. These lineages had been detected in all (AY.4.2, AY.43 and AY.98) or most (AY.120) of the local authority areas under investigation, and each had been associated with considerably more than 10,000 detections (Table 1). Table 2 summarizes the global status of these four lineages as of 9 January 2022. With the exception of the AY.43 lineage, which was prevalent in a number of European countries and associated with >267,000 detections worldwide, the majority of detections of these lineages originated from the United Kingdom.

TABLE 2 NEAR HERE

Our findings for the AY.4.2 and AY.4.3 lineages are consistent with their respective designations by the UK Health Security Agency as a distinct variant under investigation (VUI-21OCT-01) and a variant of concern [32, 33]. Preliminary investigations have indicated the AY.4.2 lineage to be associated with a higher growth rate and a higher household secondary attack rate, but with no significant reduction in vaccine effectiveness, as compared to the parent lineage [32, 34]. It remains to be established, however, whether the higher growth rate is due to enhanced transmissibility or is context-dependent [32, 35, 36]. Similarly, the status of the AY.43 lineage in terms of transmission advantage and/or immune escape remains to be determined, although further investigation is merited as new AY.43 sublineages have recently been reported from Brazil [37]. Finally, our identification of a rapid rate of spatial growth for the AY.98 and AY.120 lineages, approximating the estimated rates for AY.4.2 and AY.43, is noteworthy. Whilst very little has been documented on the epidemiological facets of these lineages, both have been identified in a number of countries in Europe and elsewhere (Table 2) and would seem to merit further investigation on the basis of the findings presented here.

With the foregoing exceptions, our analysis has shown that many emerging AY lineages in England in the spring and summer of 2021 were associated with spatial growth rates that were lower (in some instances, substantially lower) than the aggregate rate for the Delta variant (Table 1 and Figure 5). Multiple biological (e.g. reduced infectivity or

transmissibility) and contextual (e.g. progressive expansion of the national COVID-19 vaccination programme) factors may account for this observation. Importantly, there is no evidence of a temporal trend in the observed rates of spatial growth that would be suggestive of either (i) a biological selection pressure in favour of a growth advantage of emerging lineages or (ii) a progressive contextual effect in the form of, for example, increasing levels of vaccination coverage or natural immunity that would serve to retard growth rates.

Our estimates of \bar{t}_{LE} in Table 1 and Figure 5 are contingent on a range of public health interventions that operated to limit the spread of SARS-CoV-2. These interventions included: a tier system of local lockdown in October 2020; two periods of national lockdown (November–December 2020 and January–March 2021) and a phased lifting of national restrictions in the period to July 2021; a phased rollout of the national COVID-19 vaccination programme from December 2020; and the implementation of 'Plan B' control guidelines against the Omicron variant in December 2021 [38, 39]. Our results are also subject to the limitations of the available lineage data. Although the COG-UK Consortium genomic surveillance data are recognized for their extent and reliability [40], the data are formed as a sample of positive SARS-CoV-2 test results and are subject to the limitations and biases of sample data. In this context, we note that the cumulative coverage of the COG-UK records for England was estimated at 13.7% of people with positive SARS-CoV-2 test results to October 2021 [41]. Finally, our results are subject to the underpinning assumptions of the analytical procedure. In particular, the computation of \bar{t}_{LE} is dependent on the specification of the index week (i.e. the week that a given lineage was first detected in England) and the degree to which this reflects the date of actual emergence of the lineage in England. The extent to which the sample data accurately track the spatial expansion of the LE for a given lineage, the variable contributions of international travel- and community-related transmission to the development of the LE, and the geographical starting point(s) of a given lineage in the national transmission network, will also have influenced our results in unknown ways. For example, the early involvement and high degree of geographical connectivity of London and the South East may have served to accelerate the spatial transmission of the Alpha variant in the latter months of 2020 [14]. The observed rapid

spread of the Delta variant may reflect international importations and onwards transmission from multiple different geographical locations in the spring of 2021 [15, 42], whilst early cases of the Omicron variant were observed in highly connected regions at a time of reduced non-pharmaceutical interventions in November and December 2021 [43].

Further insights into the spatial growth and decay of SARS-CoV-2 lineages may be gained by application of the full swash-backwash model, but this is dependent on the substantial spatial retreat of any given lineage from the population. Here we note that, with the exception of AY.10 (last detected in July 2021) and AY.8 and Alpha (B.1.1.7 and Q) (both last detected in August/September 2021), there is evidence of the circulation of all the lineages included in Table 1 in the weeks to December 2021.

We have demonstrated, for the first time, a robust method for assessing and comparing the rate of spatial growth of multiple SARS-CoV-2 lineages in a set of geographical areas. We suggest that this approach represents a potentially valuable adjunct to routine assessments of the growth of emerging SARS-CoV-2 lineages in a defined population.

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Appendix: The COVID-19 Genomics UK (COG-UK) Consortium

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Declaration of Interest

None.

Data Availability Statement

The data that support the findings of this study are available at Wellcome Sanger Institute COVID–19 Genomic Surveillance (https://covid19.sanger.ac.uk/lineages/raw).

Table 1. Estimated rate of spatial growth (\bar{t}_{LE}) of sample SARS-CoV-2 lineages in England, September 2020–December 2021.

Variant / lineage	LTLAs (n)	Number of detections ¹	Earliest detection ² $(t=1)$	\bar{t}_{LE} (95% CI) (weeks)
Alpha (B.1.1.7 and Q)	307	153,405	39/2020 (26 Sept.)	9.90 (9.56, 10.23)
Delta (B.1.617.2 and AY)	307	722,133	12/2021 (27 March)	7.40 (7.12, 7.68)
B.1.617.2	307	17,504	13/2021 (3 April)	9.11 (8.48, 9.74)
AY.3	168	620	27/2021 (10 July)	14.12 (13.26, 14.98)
AY.4	307	547,403	12/2021 (27 March)	9.45 (9.24, 9.66)
AY.4.1	133	416	25/2021 (26 June)	9.50 (8.61, 10.40)
AY.4.2	307	77,391	25/2021 (26 June)	6.21 (5.86, 6.57)
AY.4.2.1	307	11,541	29/2021 (24 July)	9.43 (8.91, 9.94)
AY.4.3	188	781	19/2021 (15 May)	19.93 (18.92, 20.93)
AY.4.5	158	742	24/2021 (19 June)	15.06 (13.76, 16.37)
AY.5	307	26,111	15/2021 (17 April)	9.83 (9.38, 10.28)
AY.6	306	17,405	17/2021 (1 May)	8.88 (8.48, 9.27)
AY.7	276	3,627	18/2021 (8 May)	9.02 (8.39, 9.65)
AY.8	141	1,241	16/2021 (24 April)	8.60 (7.90, 9.29)
AY.9	304	10,136	14/2021 (10 April)	10.88 (10.33, 11.42)
AY.9.2	251	1,990	28/2021 (17 July)	10.44 (9.70, 11.17)
AY.10	118	683	15/2021 (17 April)	8.50 (7.96, 9.04)
AY.25	130	486	31/2021 (7 Aug.)	13.02 (12.26, 13.78)
AY.33	105	319	25/2021 (26 June)	17.84 (16.63, 19.05)
AY.34	268	3,088	30/2021 (31 July)	10.88 (10.19, 11.57)
AY.36	256	1,999	28/2021 (17 July)	12.69 (12.05, 13.32)
AY.42	115	338	27/2021 (10 July)	10.08 (8.81, 11.34)
AY.43	307	23,068	27/2021 (10 July)	5.56 (5.27, 5.86)
AY.46	305	9,461	21/2021 (29 May)	11.23 (10.72, 11.74)
AY.46.5	304	8,649	23/2021 (12 June)	10.13 (9.66, 10.61)
AY.87	172	688	20/2021 (22 May)	10.28 (9.28, 11.29)
AY.89	173	544	21/2021 (29 May)	13.97 (13.17, 14.77)
AY.90	218	1,202	21/2021 (29 May)	13.52 (12.54, 14.50)
AY.98	307	22,465	22/2021 (5 June)	6.34 (5.99, 6.70)
AY.98.1	172	621	25/2021 (26 June)	15.77 (14.80, 16.75)
AY.109	116	441	25/2021 (26 June)	17.59 (16.67, 18.52)
AY.111	296	4,649	21/2021 (29 May)	15.25 (14.46, 16.04)
AY.120	306	18,400	20/2021 (22 May)	6.56 (6.16, 6.97)
AY.122	302	5,018	21/2021 (29 May)	13.80 (13.07, 14.53)
AY.124	119	424	25/2021 (26 June)	12.99 (11.93, 14.05)
AY.125	162	534	26/2021 (3 July)	16.65 (15.79, 17.51)
Omicron (B.1.1.529 and BA)	304	23,137	47/2021 (27 Nov.) ³	$2.63 (2.56, 2.71)^4$
Average	233	46,450		11.27 (10.03, 12.50)

¹ Excludes 1,105 detections for which lineage data is suppressed and 19,550 detections for which lineage data is not available. ² Epidemiological week/year, with the last day of the week given in parentheses. ³ Excludes a lone detection in week 43 (30 October). ⁴ Indexed to week 47; \bar{t}_{LE} = 6.62 (6.53, 6.70) when indexed to week 43.

Table 2. Worldwide detection of sample Delta AY lineages with relatively high estimated rates of spatial growth (status: 9 January 2022).

Variant	Countries	Global prevalence (%)	Number of detections	Predominant countries (proportion of global detections)
AY.4.2	52	1	82,038	UK (90%), Denmark (4%), Germany (1%), Poland (1%), France (1%)
AY.43	128	4	267,426	Germany (15%), UK (13%), Denmark (13%), France (12%), Belgium (5%)
AY.98	66	1	41,507	UK (91%), USA (2%), Germany (1%), Denmark (1%), Ireland (1%)
AY.120	74	1	30,320	UK (85%), India (3%), USA (3%), Germany (2%), France (1%)

Sources: data from cov-lineages.org [27] and Latif, et al. [28–31].

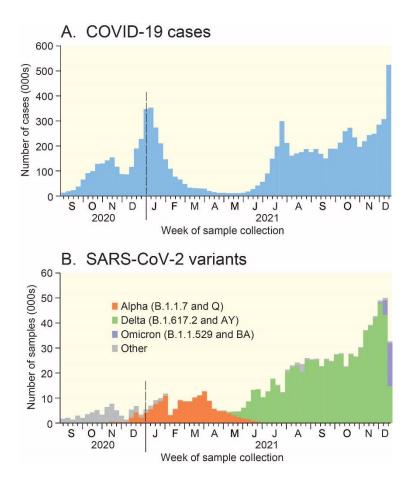


Figure 1. COVID-19 cases in England, September 2020–December 2021. (A) Positive COVID-19 test specimens as recorded by the UK Government. (B) Number of sample genomes of SARS-CoV-2 in the COG-UK database by variant to 18 December 2021. All data are plotted by week of sample collection. Sources: data from GOV.UK Coronavirus (COVID-19) in the UK [17] and COVID-19 Genomics UK (COG-UK) Consortium [18].

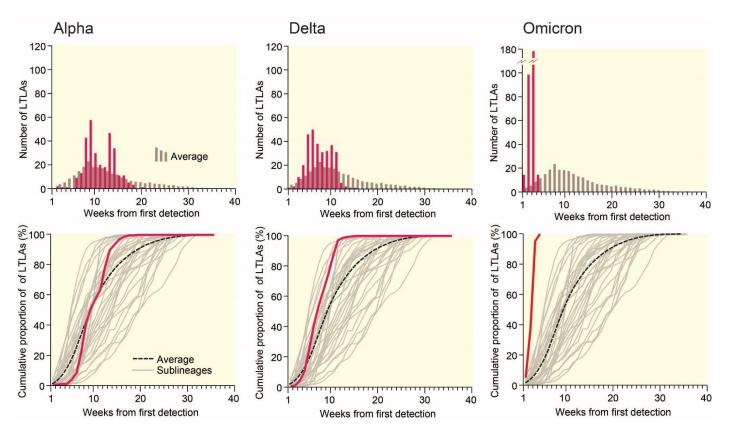


Figure 2. Spatial leading edges (*LE*) of the Alpha (B.1.1.7 and Q), Delta (B.1.617.2 and AY) and Omicron (B.1.1529 and BA) variants in England, September 2020–December 2021. The graphs plot, on a weekly basis, the non-cumulative count (upper) and cumulative proportion (lower) of LTLAs in which each of the three variants was first detected. The horizontal (time) axes are indexed to the epidemiological week of first detection (*t* = 1) of the corresponding variant. Average curves, formed across the set of sample lineages in Table 1, are plotted for reference.

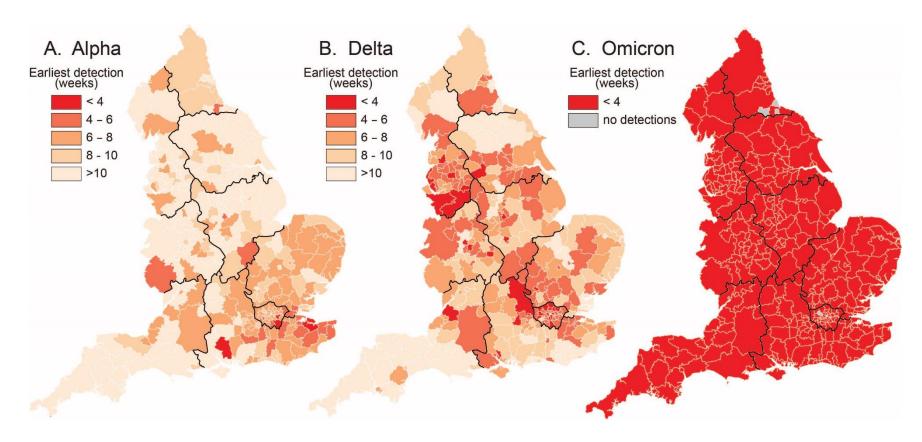


Figure 3. Spatial leading edges (*LE*) of the Alpha (B.1.1.7 and Q), Delta (B.1.617.2 and AY) and Omicron (B.1.1529 and BA) variants in the LTLAs of England, September 2020–December 2021. Maps are indexed to the epidemiological week of first detection (= week 1) of the corresponding variant and plot the number of weeks to first detection in each LTLA.

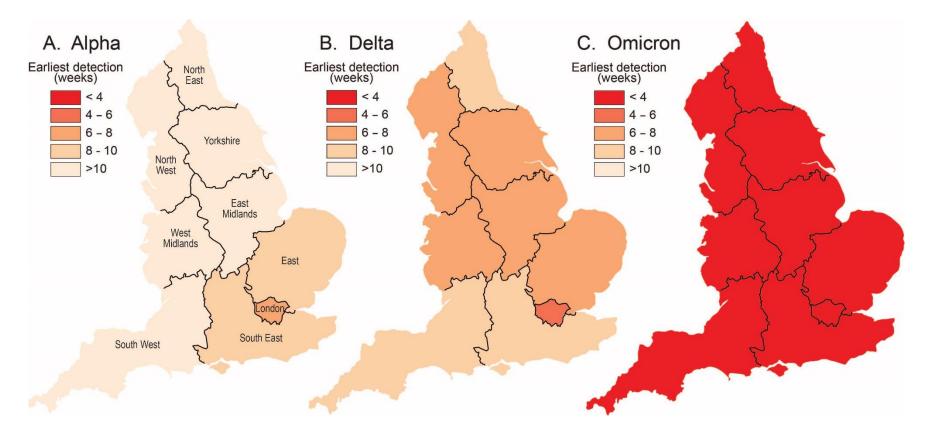


Figure 4. Spatial leadings edges (*LE*) of the Alpha (B.1.1.7 and Q), Delta (B.1.617.2 and AY) and Omicron (B.1.1529 and BA) variants in the nine standard regions of England, September 2020–December 2021. Maps plot the average time (in weeks) to first detection of a given variant in each regional subset of LTLAs.

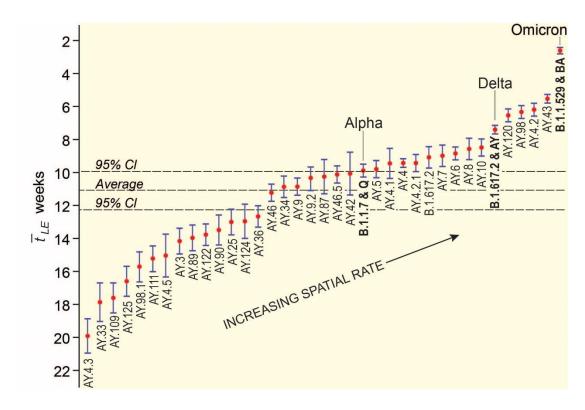


Figure 5. Estimated rate of spatial growth of sample SARS-CoV-2 lineages in England, September 2020—December 2021. The graph plots values of \bar{t}_{LE} and associated 95% CI from Table 1. Values are ordered from the lowest (left, high values of \bar{t}_{LE}) to the highest (right, low values of \bar{t}_{LE}) rates of spatial growth. Values are plotted on an inverted vertical scale to facilitate interpretation. The average value of \bar{t}_{LE} for the sample is shown for reference.

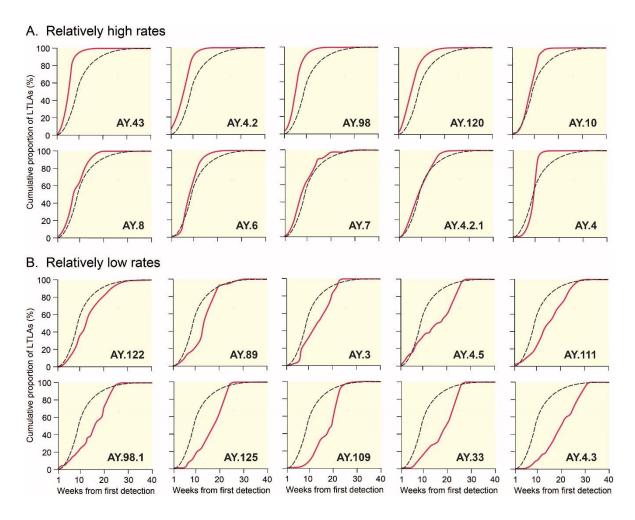


Figure 6. Spatial growth curves for sample Delta sublineages in England, March–December 2021. Curves have been formed in the manner of the lower graphs in Figure 2, with the average curve plotted for reference. Lineages are ordered according to the values of \bar{t}_{LE} in Table 1 and are defined as having relatively high (i.e. low values of \bar{t}_{LE} ; upper graphs, A) and relatively low (i.e. high values of \bar{t}_{LE} ; lower graphs, B) rates of spatial growth.