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Establishment and cryptic transmission of Zika virus in Brazil and the Americas

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1 Establishment and cryptic transmission of Zika virus in Brazil and the Americas

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93 **One Sentence Summary:** Virus genomes reveal the establishment of Zika virus in
94 Brazil and the Americas, and provide an appropriate timeframe for baseline (pre-Zika)
95 microcephaly in different regions.

96

97 **Transmission of Zika virus (ZIKV) in the Americas was first confirmed in May**
98 **2015 in northeast Brazil¹. Brazil has had the highest number of reported ZIKV**
99 **cases worldwide (more than 200,000 by 24 December 2016²) and the most cases**
100 **associated with microcephaly and other birth defects (2,366 confirmed by 31**
101 **December 2016²). Since the initial detection of ZIKV in Brazil, more than 45**
102 **countries in the Americas have reported local ZIKV transmission, with 24 of**
103 **these reporting severe ZIKV-associated disease³. However, the origin and**
104 **epidemic history of ZIKV in Brazil and the Americas remain poorly understood,**
105 **despite the value of this information for interpreting observed trends in reported**
106 **microcephaly. Here we address this issue by generating 54 complete or partial**
107 **ZIKV genomes, mostly from Brazil, and reporting data generated by a mobile**
108 **genomics laboratory that travelled across northeast Brazil in 2016. One sequence**
109 **represents the earliest confirmed ZIKV infection in Brazil. Analyses of viral**
110 **genomes with ecological and epidemiological data yield an estimate that ZIKV**
111 **was present in northeast Brazil by February 2014 and is likely to have**
112 **disseminated from there, nationally and internationally, before the first detection**
113 **of ZIKV in the Americas. Estimated dates for the international spread of ZIKV**
114 **from Brazil indicate the duration of pre-detection cryptic transmission in**
115 **recipient regions. The role of northeast Brazil in the establishment of ZIKV in**
116 **the Americas is further supported by geographic analysis of ZIKV transmission**
117 **potential and by estimates of the basic reproduction number of the virus.**
118
119

120 Previous phylogenetic analyses have indicated that the ZIKV epidemic was
121 caused by the introduction of an Asian genotype lineage into the Americas around late
122 2013, at least one year before its detection there⁴. An estimated 100 million people in
123 the Americas are predicted to be at risk of acquiring ZIKV once the epidemic has
124 reached its full extent⁵. However, little is known about the genetic diversity and
125 transmission history of the virus in Brazil⁶. Reconstructing the spread of ZIKV from
126 case reports alone is challenging because symptoms (typically fever, headache, joint
127 pain, rashes, and conjunctivitis) overlap with those caused by co-circulating
128 arthropod-borne viruses⁷ and owing to a lack of nationwide ZIKV-specific
129 surveillance in Brazil before 2016.

130 We undertook a collaborative investigation of the molecular epidemiology of
131 ZIKV in Brazil, including results from a mobile genomics laboratory that travelled
132 through northeast Brazil during June 2016 (the ZiBRA project;
133 <http://www.zibraproject.org>). Of five regions of Brazil (**Fig. 1a**), the northeast region
134 has the most notified ZIKV cases (40% of Brazilian cases) and the most confirmed
135 microcephaly cases (76% of Brazilian cases, as of 31 December 2016²), raising
136 questions about why the region has been so severely affected⁸. Furthermore, northeast
137 Brazil is the most populous region of Brazil that also has potential for year-round
138 ZIKV transmission⁹. With support from the Brazilian Ministry of Health and other
139 institutions (see **Acknowledgements**), the ZiBRA laboratory screened 1,330 samples
140 (almost exclusively serum or blood) from patients in 82 municipalities across 5
141 federal states (**Fig. 1, Extended Data Table 1a**). Samples provided by the public
142 health laboratories of each state (LACEN) and the Fundação Oswaldo Cruz
143 (FIOCRUZ) were screened for the presence of ZIKV by real-time quantitative PCR
144 (RT-qPCR).

145 On average, ZIKV viraemia persists for 10 days after infection; symptoms
146 develop after about 6 days and can last for 1–2 weeks¹⁰. In line with previous

147 observations in Colombia¹¹, we found that RT-qPCR-positive samples from northeast
148 Brazil were, on average, collected only 2 days after the onset of symptoms. The
149 median RT-qPCR cycle threshold (Ct) value of positive samples was correspondingly
150 high, at 36 (**Extended Data Fig. 1a, b**). For northeast Brazil, the time series of RT-
151 qPCR+ cases was positively correlated with the number of weekly notified cases
152 (Pearson's $r = 0.62$; **Fig. 1b**).

153 The ability of the mosquito vector *Aedes aegypti* to transmit ZIKV is
154 determined by ecological factors that affect adult survival, viral replication, and
155 infective periods¹². To investigate the receptivity of Brazilian regions to ZIKV
156 transmission we used a measure of vector climatic suitability, derived from monthly
157 temperature, relative humidity, and precipitation data¹³. Using linear regression we
158 noted that, for each Brazilian region, there is a strong association between estimated
159 climatic suitability and weekly notified cases (**Fig. 1b, c**; adjusted $R^2 > 0.84$, $P <$
160 0.001 ; **Extended Data Table 1b**). Similar to previous findings from dengue virus
161 outbreaks^{14,15}, notified ZIKV cases lag climatic suitability by about 4–6 weeks in all
162 regions, except northeast Brazil, where no time lag is evident. Despite these
163 associations, numbers of notified cases should be interpreted cautiously because co-
164 circulating dengue and chikungunya viruses exhibit symptoms similar to ZIKV, and
165 the Brazilian case reporting system has evolved through time (see **Methods**). We
166 estimated basic reproductive numbers (R_0) for ZIKV in each Brazilian region from the
167 weekly notified case data and found that R_0 was high in northeast Brazil (R_0 around 3
168 for both epidemic seasons; **Extended Data Table 1c**). Although our R_0 values are
169 approximate, in part owing to spatial variation in transmission across the large regions
170 analysed here, they are consistent with estimates from other approaches^{16,17}.

171 Encouraged by the utility of portable genomic technologies during the West
172 African Ebola virus epidemic¹⁸ we used our open protocol¹⁹ to sequence ZIKV
173 genomes directly from clinical material using MinION DNA sequencers. We were
174 able to generate virus sequences within 48h of the mobile laboratory's arrival at each
175 LACEN. In pilot experiments using a cultured ZIKV reference strain²⁰ we recovered
176 98% of the virus genome (**Extended Data Fig. 1c**). However, owing to low viral
177 copy numbers in clinical samples (**Extended Data Fig. 1a**), many sequences
178 exhibited incomplete genome coverage and required additional sequencing efforts in
179 static labs once fieldwork had been completed. Whereas average genome coverage
180 was typically high for samples with lower Ct values (85% for $Ct < 33$; **Fig. 2a**,
181 **Extended Data Table 2**), samples with higher Ct values had variable coverage (mean
182 72% for $Ct > 33$; **Fig. 2a**). Unsequenced genome regions were non-randomly
183 distributed (**Fig. 2b**), suggesting that the efficiency of PCR amplification varied
184 among primer pair combinations. We generated 36 near-complete or partial genomes
185 from the northeast, southeast and northern regions of Brazil, supplemented by nine
186 sequences from samples from Rio de Janeiro municipality. To further reconstruct Zika
187 virus transmission in the Americas, we include five new complete ZIKV genomes
188 from Colombia and four from Mexico. In addition, we append to our dataset 115
189 publicly available sequences and 85 additional genomes from ref. 21. The final
190 dataset comprised 254 ZIKV sequences, 241 of which were sampled in the Americas
191 (see **Methods**).

192 The American ZIKV epidemic comprises a single founder lineage^{4,22,23}
193 (hereafter termed Am-ZIKV) derived from Asian genotype viruses (hereafter termed
194 PreAm-ZIKV) from southeast Asia and the Pacific⁴. A sliding window analysis of
195 pairwise genetic diversity along the ZIKV genome shows that the diversity of PreAm-
196 ZIKV strains is on average about two-fold greater than that of Am-ZIKV viruses (**Fig.**

197 **2d**), reflecting a longer period of ZIKV circulation in Asia and the Pacific than in the
198 Americas. The genetic diversity of Am-ZIKV strains will increase in the future and
199 updated diagnostic assays are recommended to guarantee RT-qPCR sensitivity²⁴.

200 It has been suggested that recent ZIKV epidemics may be linked causally to a
201 higher apparent evolutionary rate for the Asian genotype than the African
202 genotype^{25,26}. However, such comparisons are confounded by an inverse relationship
203 between the timescale of observation and estimated evolutionary rates²⁷. Regression
204 of sequence sampling dates against root-to-tip genetic distances indicates that
205 molecular clock models can be applied reliably to the Asian ZIKV lineage (**Fig. 2c**,
206 **Extended Data Figs 2, 3**). We estimate the whole-genome evolutionary rate of Asian
207 ZIKV to be 1.12×10^{-3} substitutions per site per year (95% Bayesian credible interval
208 (BCI) $0.97\text{--}1.27 \times 10^{-3}$), consistent with other estimates for this lineage^{4,26}. We found
209 no significant differences in evolutionary rates among ZIKV genome regions
210 (**Extended Data Table 3a**). The estimated ratio of divergence at nonsynonymous and
211 synonymous sites (dN/dS) of the Am-ZIKV lineage is low (0.11, 95% confidence
212 interval 0.10–0.13), as observed for other vector-borne flaviviruses²⁸, but is higher
213 than that of PreAm-ZIKV viruses (0.061, 0.047–0.077), probably owing to the raised
214 probability of observing slightly deleterious changes in short-term datasets, as
215 observed during previous epidemics²⁹.

216 We used two phylogeographic approaches with different assumptions^{30,31} to
217 reconstruct the origins and spread of ZIKV in Brazil and the Americas. We dated the
218 common ancestor of ZIKV in the Americas (node B, **Fig. 3**) to Jan 2014 (95% BCI
219 October 2013–April 2014; **Extended Data Tables 3b, c**), in line with previous
220 estimates^{4,26}. We find evidence that northeast Brazil played a central role in the
221 establishment and dissemination of Am-ZIKV. Although northeast Brazil is the most
222 probable location of node B (location posterior support 0.83, **Fig. 3**), the current data
223 do not allow us to exclude the hypothesis that node B was in the Caribbean (**Fig. 3**
224 dashed branches) owing to the presence of two sequences from Haiti in one of its
225 descendant lineages. More importantly, most Am-ZIKV sequences descend from a
226 radiation of lineages (node C and its immediate descendants; **Fig. 3**) dated to late
227 February 2014 (95% BCIs of node C, November 2013–May 2014). Node C is more
228 strongly inferred to have existed in northeast Brazil (location posterior support 0.99,
229 **Fig. 3**). All 20 replicate analyses performed on subsampled datasets place node C in
230 Brazil, and 14 of them place node C in northeast Brazil (**Extended Data Fig. 4**).
231 Consequently, we conclude that node C reflects the crucial turning point in the
232 emergence of ZIKV in the Americas. If further data show that node B did exist in
233 Haiti, then it is likely that Haiti acted as an intermediate ‘stepping stone’ for the
234 arrival and establishment of Am-ZIKV in Brazil, from where the virus subsequently
235 spread to other regions. This perspective is consistent with the lower population size
236 of Haiti compared to Brazil. We infer that node C was present in northeast Brazil
237 several months before three notable events, each of which also occurred in northeast
238 Brazil: (i) the retrospective identification of a cluster of suspected but unconfirmed
239 ZIKV cases in December 2014¹; (ii) the collection of the oldest ZIKV genome
240 sequence from Brazil, reported here, sampled in February 2015; and (iii) confirmation
241 of cases of ZIKV transmission in northeast Brazil in March 2015^{32,33}.

242 Our results further indicate that viruses from northeast Brazil were important
243 for the continental spread of ZIKV. Within Brazil, we find instances of virus lineage
244 movement from northeast to southeast Brazil; most of these events are dated to the
245 second half of 2014 and led to onwards transmission in Rio de Janeiro (RJ1–RJ4; **Fig.**
246 **3**) and São Paulo states (SP1; **Fig. 3**). We infer that ZIKV lineages disseminated from

247 northeast Brazil to elsewhere in Central America, the Caribbean, and South America.
248 Most Am-ZIKV strains sampled outside Brazil fall into four well-supported
249 phylogenetic groups (**Fig. 3**); three (SA1/CB1, CA1 and SA2) are inferred to have
250 been exported from northeast Brazil between July 2014 and April 2015, whereas the
251 Caribbean clade CB2 appears to have originated from southeast Brazil around March
252 2015 (**Figs 3, 4**). Each viral lineage export occurred during a period of climatic
253 suitability for vector transmission in the recipient location (**Fig. 4**). For the earliest
254 exports to Central America (CA1) and South America (SA1), there is an estimated
255 11–12-month gap between the date of export and the date of ZIKV detection in the
256 recipient location, suggesting a complete season of undetected transmission. These
257 periods of cryptic transmission are relevant to studies of spatiotemporal trends in
258 reported microcephaly, because they help to define the appropriate timeframe for
259 baseline (pre-ZIKV) microcephaly in each region.

260 Large-scale surveillance of ZIKV is challenging because many cases may be
261 asymptomatic, and ZIKV co-circulates in some regions with other arthropod-borne
262 viruses that have overlapping symptoms (for example, dengue, chikungunya, Mayaro,
263 and Oropouche viruses). However combining virus genomic and epidemiological data
264 can generate insights into vector-borne virus transmission. A system of continuous
265 and structured virus sequencing in Brazil, integrated with surveillance data, could
266 provide timely information to inform effective responses against Zika and other
267 viruses, including the recently re-emerged yellow fever virus³⁴.

268

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367
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416 MRTN, and LCJA undertook fieldwork and experiments. NRF, JT, C-HW, OGP, JR
417 and LdP performed genetic analyses. NRF, MUG, OGP and SC performed
418 epidemiological analyses. NRF, JQ, MUGK, NL and OGP wrote the manuscript.
419 ECH, AR, TB, MRTN, ECS and LCJA edited the manuscript. Other authors were
420 critical for coordination, collection, processing, sequencing and bioinformatics of
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433

434

435 **Fig. 1. Geographic and temporal distribution of ZIKV in Brazil.** **a.** Sampling
436 location of genome sequences from Brazil and the Americas. Federal states in Brazil
437 are coloured according to 5 geographic regions (lower inset). A red line surrounds the
438 states surveyed by the ZiBRA mobile lab in 2016. State codes are PA=Pará,
439 MA=Maranhão, CE=Ceará, TO=Tocantins, RN=Rio Grande do Norte, PB=Paraíba,
440 PE=Pernambuco, AL=Alagoas, BA=Bahia, RJ=Rio de Janeiro, SP=São Paulo.
441 Underlined states represent those from which sequences in this study were generated
442 (upper inset). Publicly available sequences were also collated from non-underlined
443 states. **b.** Confirmed and notified ZIKV cases in NE Brazil. Upper panel shows the
444 temporal distribution of RT-qPCR+ cases detected during ZiBRA fieldwork. Only
445 samples with known collection dates are included (n=138 out of 181 confirmed
446 cases). Lower panel shows notified ZIKV cases in NE Brazil between 01 Jan 2015
447 and 19 Nov 2016 (n=122,779). The dashed line represents the average climatic vector
448 suitability score for NE Brazil (**Methods**). The vertical arrow indicates date of ZIKV
449 confirmation in NE Brazil/Americas¹. **c.** Notified ZIKV cases in the Centre-West,
450 Southeast, North, and South regions of Brazil (clockwise from top left). The dashed
451 lines represent the average climatic vector suitability score for each region.
452

453 **Fig. 2. Zika virus genetic diversity and sequencing statistics.** **a.** The percentage of
454 ZIKV genome sequenced plotted against RT-qPCR Ct-value, for each sample. Each
455 circle represents a sequence recovered from an infected individual in Brazil and is
456 coloured by sampling location. **b.** Illustration of sequencing coverage across the
457 ZIKV genome for the ZiBRA sequences, including data generated by both mobile and
458 static laboratories. **c.** Regression of sequence sampling dates against root-to-tip
459 genetic distances in a maximum likelihood phylogeny of the Asian-ZIKV lineage.
460 **Extended Data Fig. 2b** contains a comparable analysis that also includes P6-740 (the
461 oldest Asian-ZIKV strain collected in 1966). **d.** Average pairwise genetic diversity of
462 the PreAm-ZIKV strains (grey line) and of the Am-ZIKV lineage (black line),
463 calculated using a sliding window of 300 nucleotides with a step size of 50
464 nucleotides.

465
466 **Fig. 3. Phylogeography of ZIKV in the Americas.** Maximum clade credibility
467 phylogeny, estimated from complete and partial Am-ZIKV genomes using a
468 molecular clock phylogeographic approach (**Methods**). Terminal branches with
469 yellow circles indicate sequences reported in this study. Terminal branches with no
470 circles and reduced opacity are those reported in a companion paper²⁰. Thin vertical
471 grey boxes indicate statistical uncertainty of estimated dates of nodes A, B and C
472 (**Extended Data Table 3c**). Branch colours indicate the most probable ancestral
473 lineage locations. Diamonds at internal nodes are sized in proportion to clade
474 posterior probabilities. For selected nodes, coloured numbers show the posterior
475 probabilities of ancestral locations and numbers in grey are clade posterior
476 probabilities. Asterisks indicate the three available genomes from microcephaly cases.
477 A black arrow indicates the oldest Brazilian ZIKV sequence. The grey arrow and
478 dotted line denotes when ZIKV was first confirmed in the Americas¹. Nodes A and B
479 are equivalent to the nodes named identically in⁴. Text labels along the bottom of the
480 figure denote clades of sequences from regions outside of NE Brazil. RJ1 to RJ4 are
481 clades from Rio de Janeiro state, TO from Tocantins, and SP1 from São Paulo state.
482 Clades from outside Brazil are denoted CB1 and CB2 (Caribbean), SA1 and SA2

483 (South America excluding Brazil), and CA1 (Central America). Thin grey horizontal
484 lines along the bottom of the figure denote sequences from Brazil.

485

486 **Fig. 4. Establishment of Am-ZIKV in the Americas.** The earliest inferred dates of
487 lineage export to non-Brazilian regions, represented by box-and-whisker plots. Each
488 plot corresponds to the earliest movement between a pair of locations with well-
489 supported virus lineage migration. The first exports to South America outside Brazil
490 (SA1 in **Fig. 3**), to Central America (CA1) and to the Caribbean (CB1) are shown in
491 panels **a-c**, respectively. Box and whisker plots were generated in ggplot2, with boxes
492 representing the median and interquartile ranges of the estimated date of earliest
493 movement. In each of **a-c**, dashed lines show the estimated climatic vector suitability
494 score for each recipient region, averaged across the countries for which sequence data
495 is available (see **Methods**). In each of **a-c**, the bar plots show available notified ZIKV
496 case data (plots adapted from PAHO) for the countries with the earliest confirmed
497 cases (Colombia⁶¹ in panel **a**, Mexico⁶² in **b**, and Puerto Rico⁶³ in **c**). Coloured arrows
498 indicate the earliest confirmation of ZIKV autochthonous cases in each non-Brazilian
499 region. The vertical dashed line represents the date of ZIKV confirmation in the
500 Americas.

501

Figure 1

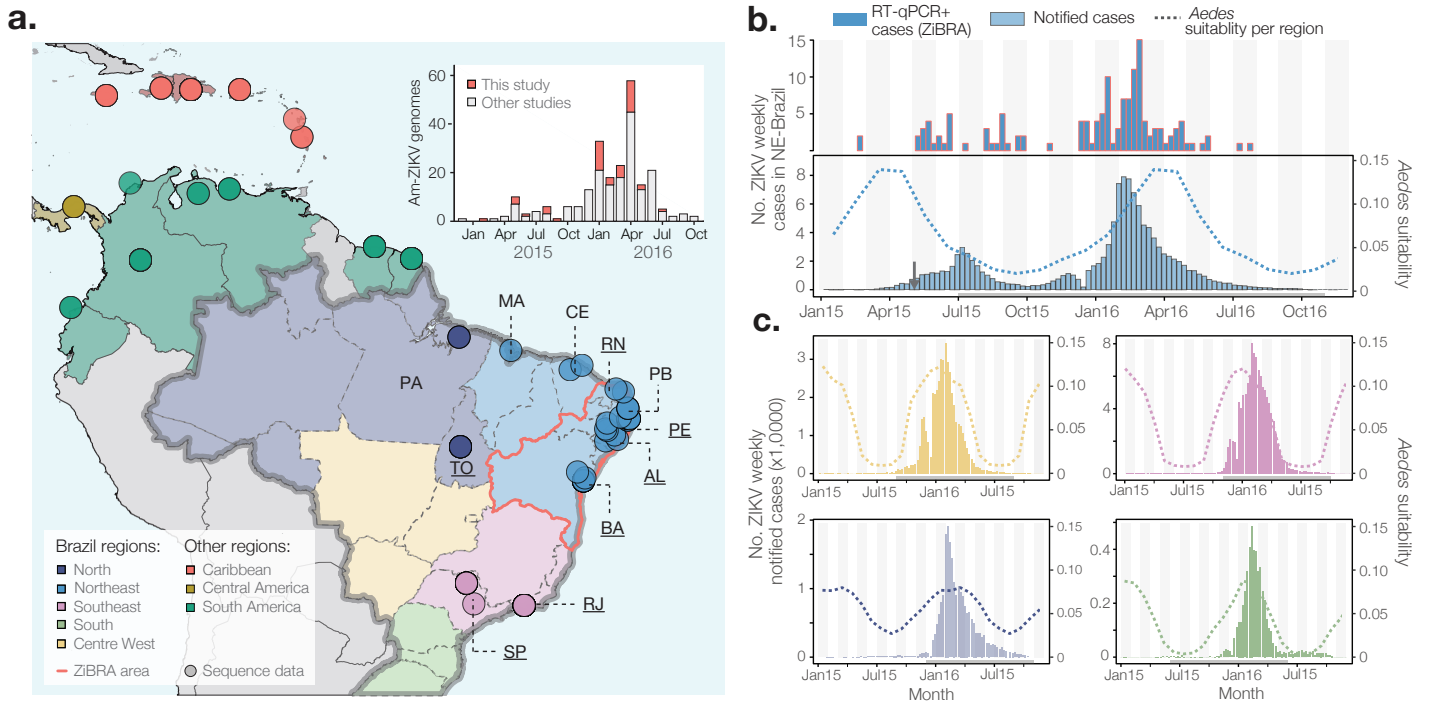


Figure 2

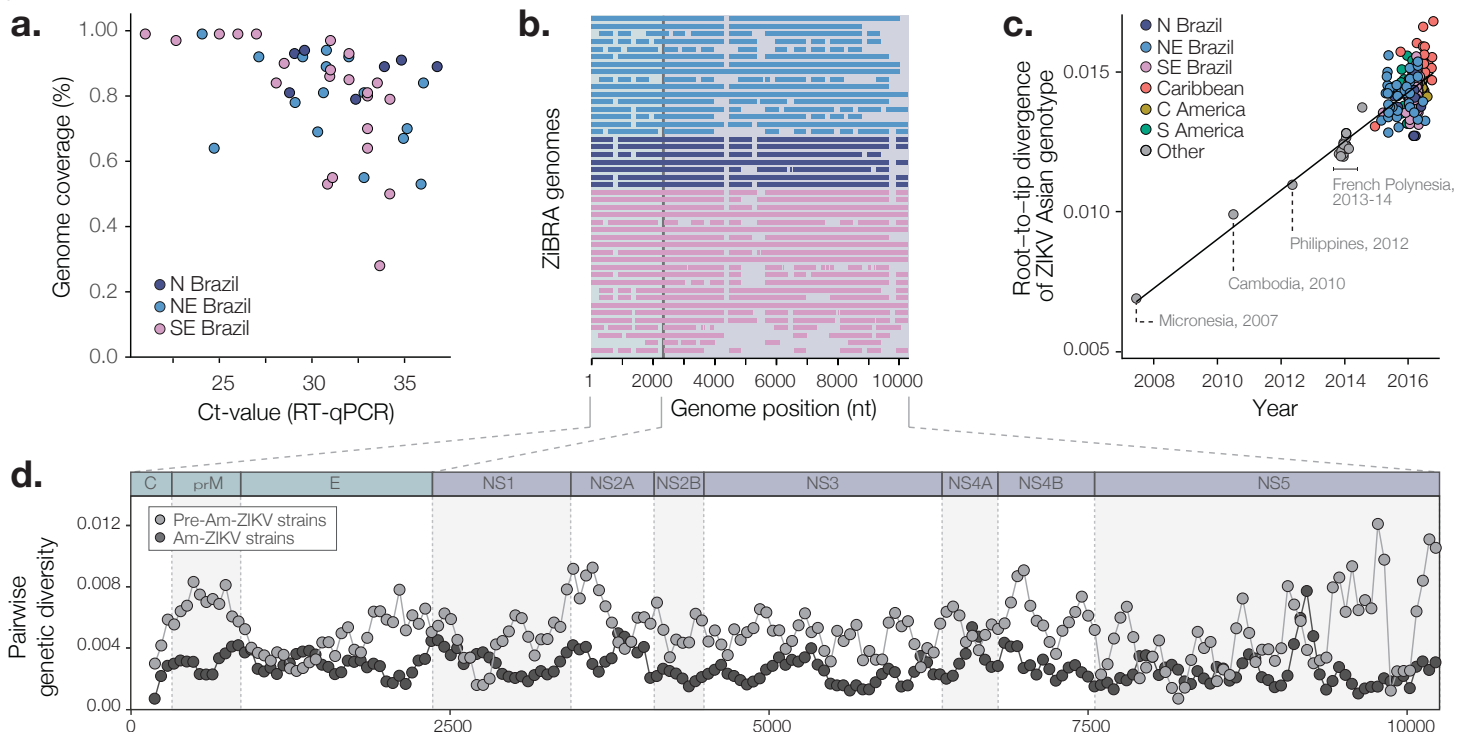


Figure 3

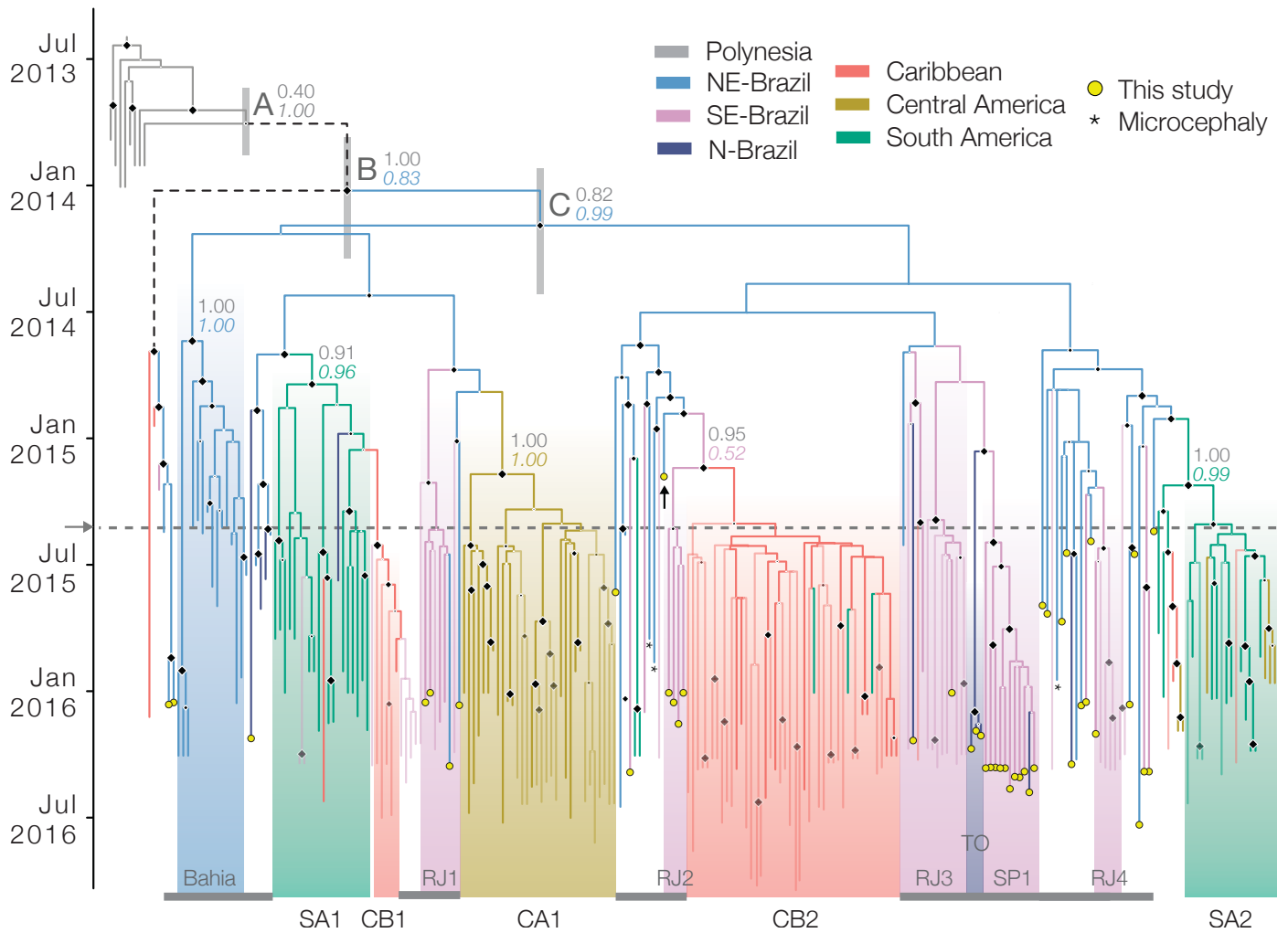
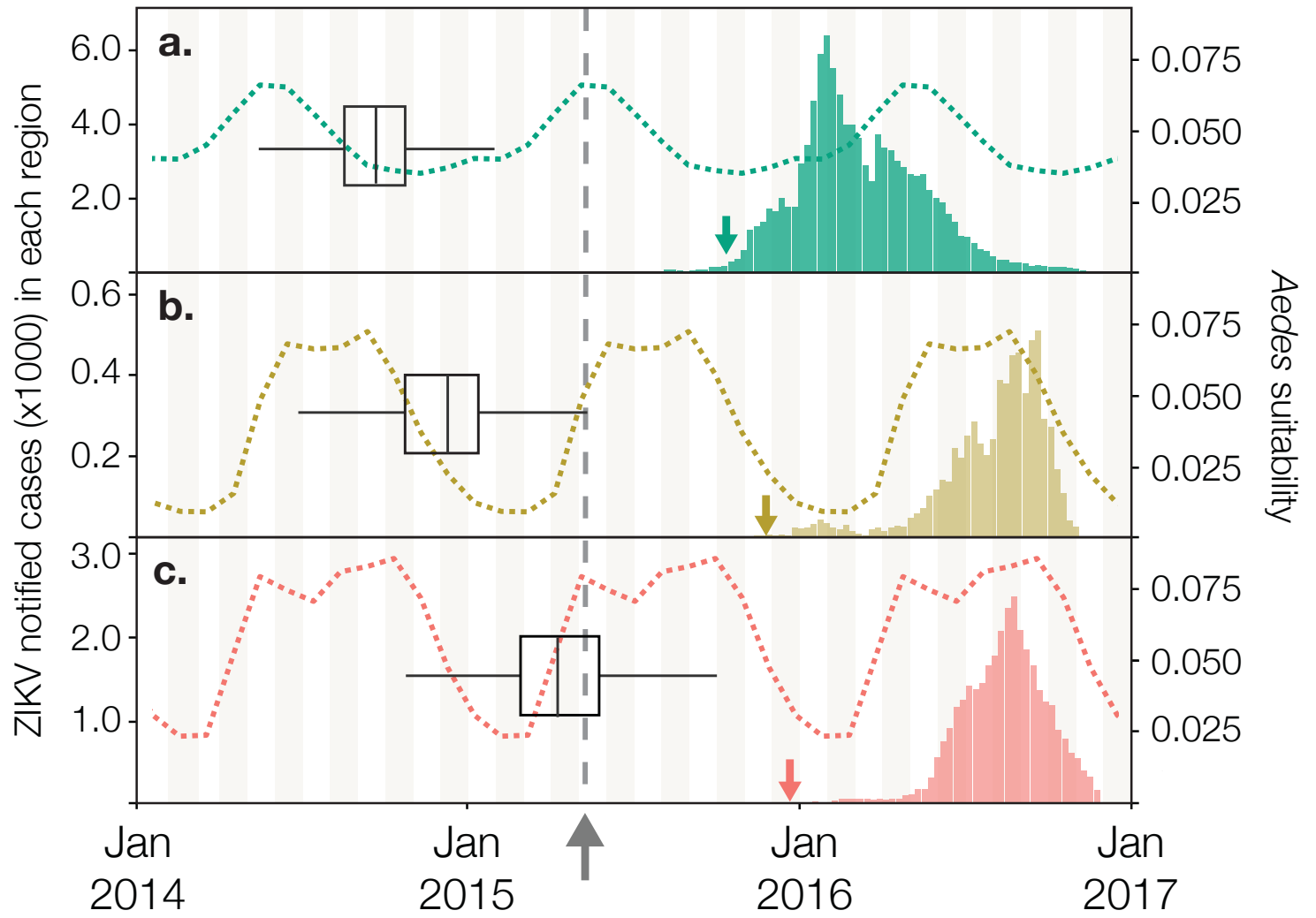


Figure 4



502 **Methods**

503 **Sample collection**

504 Between the 1st and 18th June 2016, 1330 samples from cases notified as ZIKV
505 infected were tested for ZIKV infection in the Northeast region of Brazil (NE Brazil).
506 During this period, 4 of the 5 laboratories in the region visited by the ZiBRA project
507 were in the process of implementing molecular diagnostics for ZIKV. The ZiBRA
508 team spent 2-3 days in each state central public health laboratory (LACEN). The
509 samples analysed had been previously collected from patients who had attended a
510 municipal or state public health facility, presenting maculopapular rash and at least
511 two of the following symptoms: fever, conjunctivitis, polyarthralgia, or periarticular
512 edema. The majority of samples were linked to a digital record that collated
513 epidemiological and clinical data: date of sample collection, location of residence,
514 demographic characteristics, and date of onset of clinical symptoms (when available).

515 The ZiBRA project was supported by the Brazilian Ministry of Health (MoH) as part
516 of the emergency public health response to Zika. Samples had been previously
517 obtained for routine diagnostic purposes from persons visiting local clinics by the
518 Brazilian National Health Surveillance network as part of Zika virus surveillance
519 activities. In these cases, we used samples without informed consent with the approval
520 of the Brazilian Ministry of Health. Specifically, residual anonymized clinical
521 diagnostic samples, with no or minimal risk to patients, were provided for research
522 and surveillance purposes within the terms of Resolution 510/2016 of CONEP
523 (Comissão Nacional de Ética em Pesquisa, Ministério da Saúde; National Ethical
524 Committee for Research, Ministry of Health). For samples obtained from patients
525 engaged in longitudinal studies of Zika virus in São Paulo and Tocantins states,
526 informed consent was obtained (IRB CAAE 53153916.7.0000.0065). Samples from
527 patients followed in Salvador and Feira de Santana were analysed under institutional
528 approval from CPqGM/FioCruz/BA (1.184.454). Urine and plasma samples from Rio
529 de Janeiro were obtained from patients at the Fiocruz Viral Hepatitis Ambulatory
530 (Oswaldo Cruz Institute, Rio de Janeiro, Brazil) with Institutional Review Board
531 approval (IRB142/01) from the Oswaldo Cruz Institute. RNA was extracted at the
532 Paul-Ehrlich-Institut and sequenced at the University of Birmingham, UK.

533

534 **Nucleic acid isolation and RT-qPCR**

535 Serum, blood and urine samples were obtained from patients 0 to 228 days after first
536 symptoms (**Extended Data Table 1a**). Viral RNA was isolated from 200 µl Zika-
537 suspected samples using either the NucliSENS easyMag system (BioMerieux,
538 Basingstoke, UK) (Ribeirão Preto samples), the ExiPrep Dx Viral RNA Kit
539 (BIONEER, Republic of Korea) (Rio de Janeiro samples) or the QIAamp Viral RNA
540 Mini kit (QIAGEN, Hilden, Germany) (all other samples) according to the
541 manufacturer's instructions. Ct values were determined for all samples by probe-
542 based RT-qPCR against the *prM* target (using 5'FAM as the probe reporter dye) as
543 previously described³⁴. RT-qPCR assays were performed using the QuantiNova Probe
544 RT-qPCR Kit (20 µl reaction volume; QIAGEN) with amplification in the Rotor-
545 Gene Q (QIAGEN) following the manufacturer's protocol. Primers/probe were
546 synthesised by Integrated DNA Technologies (Leuven, Belgium). The following
547 reaction conditions were used: reverse transcription (50°C, 10 min), reverse
548 transcriptase inactivation and DNA polymerase activation (95°C, 20 sec), followed by
549 40 cycles of DNA denaturation (95°C, 10 secs) and annealing-extension (60°C, 40

550 sec). Positive and negative controls were included in each batch; however, due to the
551 large number of samples tested in a short time it was possible only to run each sample
552 without replication.

553

554 **Whole genome sequencing**

555 Sequencing was attempted on all positive samples obtained from NE Brazil regardless
556 of Ct value. All samples collected in Brazil that are reported in this study were
557 sequenced with the Oxford Nanopore MinION. Sequencing statistics can be found in
558 **Extended Data Table 2**. The protocol employed cDNA synthesis with random
559 primers followed by gene specific multiplex PCR and is presented in detail in Quick
560 et al.¹⁸. In brief, extracted RNA was converted to cDNA using the Protoscript II First
561 Strand cDNA synthesis Kit (New England Biolabs, Hitchin, UK) and random
562 hexamer priming. ZIKV genome amplification by multiplex PCR was attempted
563 using the ZikaAsianV1 primer scheme and 40 cycles of PCR using Q5 High-Fidelity
564 DNA polymerase (NEB) as described in Quick et al.¹⁸. PCR products were cleaned-up
565 using AmpureXP purification beads (Beckman Coulter, High Wycombe, UK) and
566 quantified using fluorimetry with the Qubit dsDNA High Sensitivity assay on the
567 Qubit 3.0 instrument (Life Technologies). PCR products for samples yielding
568 sufficient material were barcoded and pooled in an equimolar fashion using the Native
569 Barcoding Kit (Oxford Nanopore Technologies, Oxford, UK). Sequencing libraries
570 were generated from the barcoded products using the Genomic DNA Sequencing Kit
571 SQK-MAP007/SQK-LSK208 (Oxford Nanopore Technologies). Sequencing libraries
572 were loaded onto a R9/R9.4 flowcell and data was collected for up to 48 hours but
573 generally less. As described¹⁸, consensus genome sequences were produced by
574 alignment of two-direction reads to a Zika virus reference genome (strain H/PF/2013,
575 GenBank Accession number: KJ776791) followed by nanopore signal-level detection
576 of single nucleotide variants. Only positions with $\geq 20x$ genome coverage were used to
577 produce consensus alleles. Regions with lower coverage, and those in primer-binding
578 regions were masked with N characters. Validation of our sequencing approach on the
579 MinION platform was undertaken by using the MinION platform to sequence a WHO
580 reference strain of Zika virus that was also sequenced using the Illumina Miseq
581 platform¹⁹; identical consensus sequences were recovered regardless of the MinION
582 chemistry version employed (R7.3, R9 and R9.4) (**Extended Data Fig. 1c**).

583

584 **Collation of genome-wide data sets**

585 Our complete and partial genome sequences were appended to a global data set of all
586 available published ZIKV genome sequences (up until January 2017) using an in-
587 house script that retrieves updated GenBank sequences on a daily basis. In addition to
588 the genomes generated from samples collected in NE Brazil during ZiBRA fieldwork,
589 samples were sent directly to University of São Paulo and elsewhere for sequencing.
590 Thirteen genomes from Ribeirão Preto, São Paulo state (SP; SE-Brazil region) and
591 seven genomes from Tocantins (TO; N-Brazil region) were sequenced at University
592 of São Paulo. Nine genomes from Rio de Janeiro (RJ; SE-Brazil region) were
593 sequenced in Birmingham, UK, and added to our dataset. All these genomes were
594 generated using the same primer scheme as the ZiBRA samples collected in NE
595 Brazil¹⁸. In addition to these 45 sequences from Brazil, we further included in analysis
596 9 genomes from ZIKV strains sampled outside of Brazil in order to contextualise the

597 genetic diversity of Brazilian ZIKV, giving rise to a final data set of 54 sequences.
598 Specifically, we included 5 genomes from samples collected in Colombia and 4 new
599 genomes from Mexico, which were generated using the protocols described in refs. ³⁵
600 and ²², respectively.

601 GenBank sequences belonging to the African genotype of ZIKV were identified using
602 the Arboviral genotyping tool (<http://bioafrica2.mrc.ac.za/regagenotype/typingtool/aedesviruses>) and excluded from subsequent analyses, as our
603 focus of study was the Asian genotype of ZIKV, and the Am-ZIKV lineage in
604 particular. To assess the robustness of molecular clock dating estimates to the
605 inclusion of older sequences, analyses were performed both with and without the P6-
606 740 strain, the oldest known strain of the ZIKV-Asian genotype (sampled in 1966 in
607 Malaysia). Our final alignment comprised the sequences reported in this study ($n=54$)
608 plus publicly available ZIKV-Asian genotype sequences, as of 1st March 2017
609 ($n=115$). We also included in our analysis 85 additional genomes from a companion
610 paper²⁰. The dataset used for analysis therefore included sequences from 254 Zika
611 virus isolates, 241 of which were from the Americas. Unpublished but publicly
612 available genomes were included in our analysis only if we had written permission
613 from those who generated the data (see **Acknowledgments**).

615

616 **Maximum likelihood analysis and recombination screening**

617 Preliminary maximum likelihood (ML) trees were estimated with ExaMLv3³⁶ using a
618 per-site rate category model and a gamma distribution of among site rate variation.
619 For the final analyses, ML trees were estimated using PhyML³⁷ under a GTR
620 nucleotide substitution model³⁸, with a gamma distribution of among site rate
621 variation, as selected by jModeltest.v.2³⁹. Branch support was inferred using 100
622 bootstrap replicates³⁷. Final ML trees were estimated with NNI and SPR heuristic tree
623 search algorithms; equilibrium nucleotide frequencies and substitution model
624 parameters were estimated using ML³⁷ (see **Extended Data Fig. 3**).

625 Recombination may impact evolutionary estimates⁴⁰ and has been shown to be
626 present in the ZIKV-African genotype⁴¹. In addition to restricting our analysis to the
627 Asian genotype of ZIKV, we employed the 12 recombination detection methods
628 available in RDPv4⁴² and the Phi-test approach⁴³ available in SplitsTree⁴⁴ to further
629 search for evidence of recombination in the ZIKV-Asian lineage. No evidence of
630 recombination was found.

631 Analysis of the temporal molecular evolutionary signal in our ZIKV alignments was
632 conducted using TempEst⁴⁵. In brief, collection dates in the format yyyy-mm-dd (ISO
633 8601 standard) were regressed against root-to-tip genetic distances obtained from the
634 ML phylogeny. When precise sampling dates were not available, a precision of 1
635 month or 1 year in the collection dates was taken into account.

636 To compare the pairwise genetic diversity of PreAm-ZIKV strains from Asia and the
637 Pacific with Am-ZIKV viruses from the Americas, we used a sliding window
638 approach with 300 nt wide windows and a step size of 50 nt. Sequence gaps were
639 ignored; hence the average pairwise difference per window was obtained by dividing
640 the total pairwise nucleotide differences by the total number of pairwise comparisons.

641

642

643 **Molecular clock phylogenetics and gene-specific d_N/d_S estimation**

644 To estimate Bayesian molecular clock phylogenies, analyses were run in duplicate
645 using BEASTv.1.8.4⁴⁶ for 30 million MCMC steps, sampling parameters and trees
646 every 3000 steps. We employed a model selection procedure using both path-
647 sampling and stepping stone models⁴⁷ to estimate the most appropriate combination of
648 molecular clock and coalescent models for Bayesian phylogenetic analysis. The best
649 fitting combination was a Bayesian skyline tree prior and a relaxed molecular clock
650 model, with log-normally distributed variation in rates among branches (**Extended**
651 **Data Table 3b**). A non-informative continuous time Markov chain reference prior⁴⁹
652 on the molecular clock rate was used. Convergence of MCMC chains was checked
653 with Tracer v.1.6. After removal of burn-in, posterior tree distributions were
654 combined and subsampled to generate an empirical distribution of 1,500 molecular
655 clock trees.

656 To estimate rates of evolution per gene we partitioned the alignment into 10 genes (3
657 structural genes *C*, *prM*, *E*, and 7 non-structural genes *NS1*, *NS2A*, *NS2B*, *NS3*, *NS4A*,
658 *NS4B* and *NS5*) and employed a SDR06 substitution model⁴⁸ and a strict molecular
659 clock model, using an empirical distribution of molecular clock phylogenies. To
660 estimate the ratio of nonsynonymous to synonymous substitutions per site (d_N/d_S) for
661 the PreAm-ZIKV and the Am-ZIKV lineages, we used the single likelihood ancestor
662 counting (SLAC) method⁵⁰ implemented in HyPhy⁵¹. This method was applied to two
663 distinct codon-based alignments and their corresponding ML trees which comprised
664 the PreAm-ZIKV and Am-ZIKV sequences, respectively.

665

666 **Phylogeographic analysis**

667 We investigated virus lineage movements using our empirical distribution of
668 phylogenetic trees and the sampling location of each ZIKV sequence. The sampling
669 location of sequences collected from returning travellers was set to the travel
670 destination in the Americas where infection likely occurred. We discretised sequence
671 sampling locations in Brazil into the geographic regions defined in the main text. The
672 number of sequences per region available for analysis was 10 for N Brazil, 41 for NE
673 Brazil and 54 for SE Brazil. No viral genetic data was available for the Centre-West
674 (CW) and the South (S) Brazilian regions. We similarly discretised the locations of
675 ZIKV sequences sampled outside of Brazil. These were grouped according to the
676 United Nations M49 coding classification of macro-geographical regions. Our
677 analysis included 53 sequences from the Caribbean, 38 from Central America, 17
678 from Polynesia, 37 from South America (excluding Brazil), 3 from Southeast Asia
679 and 1 from Micronesia. To account for the possibility of sampling bias arising from a
680 larger number of sequences from particular locations, we repeated all
681 phylogeographic analyses using (i) the full dataset ($n=254$) and (ii) ten jackknife
682 resampled datasets ($n=74$) in which taxa from each location (except for Southeast
683 Asia and Micronesia) were randomly sub-sampled to 10 sequences (the number of
684 sequences available for N-Brazil).

685 Phylogeographic reconstructions were conducted using two approaches; (i) using the
686 asymmetric⁵² discrete trait evolution models implemented in BEASTv1.8.4⁴⁶ and (ii)
687 using the Bayesian structured coalescent approximation (BASTA)²⁹ implemented in
688 BEAST2v.2. The latter has been suggested to be less sensitive to sampling biases⁵³.
689 For both approaches, maximum clade credibility trees were summarized from the

690 MCMC samples using TreeAnnotator after discarding 10% as burn-in. The posterior
691 estimates of the location of nodes A, B and C (depicted in **Fig. 3**) from these two
692 analytical approaches (applied to both the complete and jackknifed data sets) can be
693 found in **Extended Data Fig. 4**.

694 For the discrete trait evolution approach, we counted the expected number of
695 transitions among each pair of locations (net migration) using the robust counting
696 approach^{54,55} available in BEASTv1.8.4⁴⁶. We then used those inferred transitions to
697 identify the earliest estimated ZIKV introductions into new regions. These viral
698 lineage movement events were statistically supported (with Bayes factors > 3) using
699 the BSSVS (Bayesian stochastic search variable selection) approach implemented in
700 BEASTv.1.8.4³⁰. Box plots for node ages were generated using the ggplot2⁵⁶ package
701 in R software⁵⁷.

702

703 **Epidemiological analysis**

704 Weekly suspected ZIKV data per Brazilian region were obtained from the Brazilian
705 Ministry of Health (MoH). Cases were defined as suspected ZIKV infection when
706 patients presented maculopapular rash and at least two of the following symptoms:
707 fever, conjunctivitis, polyarthralgia or periarticular edema. Because notified suspected
708 ZIKV cases are based on symptoms and not molecular diagnosis, it is possible that
709 some notified cases represent other co-circulating viruses with related symptoms,
710 such as dengue and Chikungunya viruses. Further, case reporting may have varied
711 among regions and through time. Data from 2015 came from the pre-existing MoH
712 sentinel surveillance system that comprised 150 reporting units throughout Brazil,
713 which was eventually standardised in Feb 2016 in response to the ZIKV epidemic.
714 We suggest that these limitations should be borne in mind when interpreting the
715 ZIKV notified case data and we consider the R_0 values estimated here to be
716 approximate. That said, our time series of RT-qPCR+ ZIKV diagnoses from NE
717 Brazil qualitatively match the time series of notified ZIKV cases from the same
718 region (**Fig. 1b**). To estimate the exponential growth rate of the ZIKV outbreak in
719 Brazil, we fit a simple exponential growth rate model to each stage of the weekly
720 number of suspected ZIKV cases from each region separately:

721

$$722 \quad I_w = I_0 \exp(r_w \cdot w) \quad (1)$$

723

724 where I_w is the number of cases in week w . As described in main text, the Brazilian
725 regions considered here were NE Brazil, N-Brazil, S-Brazil, SE-Brazil, and CW-
726 Brazil. The time period over which exponential growth occurs was determined by
727 plotting the log of I_w and selecting the period of linearity (**Extended Data Fig. 5**). A
728 linear model was then fitted to this period to estimate the weekly exponential growth
729 rate r_w :

730

$$731 \quad \ln(I_w) = \ln(I_0) + r_w \cdot w \quad (2)$$

732

733 Let $g(\cdot)$ be the probability density distribution of the epidemic generation time (i.e.
734 the duration between the time of infection of a case and the mean time of infection of

735 its secondary infections). The following formula can be used to derive the
736 reproduction number R from the exponential growth rate r and density $g(\cdot)$ ⁵⁸.

737

$$738 \quad R = \frac{1}{\int_0^{\infty} \exp(-r.t)g(t)dt} \quad (3)$$

739 In our baseline analysis, following Ferguson et al.⁵⁹ we assume that the ZIKV
740 generation time is Gamma-distributed with a mean of 20.0 days and a standard
741 deviation (SD) of 7.4 days. In a sensitivity analysis, we also explored scenarios with
742 shorter mean generation times (10.0 and 15.0 days) but unchanged coefficient of
743 variation $SD/mean=7.4/20=0.37$ (**Extended Data Table 1c**).

744

745 **Association between *Aedes aegypti* climatic suitability and ZIKV notified cases**

746 To account for seasonal variation in the geographical distribution of the ZIKV vector
747 *Aedes aegypti* in Brazil we fitted high-resolution maps⁶⁰ to monthly covariate data.
748 Covariate data included time-varying variables, such as temperature-persistence
749 suitability, relative humidity, and precipitation, as well as static covariates such as
750 urban versus rural land use. Maps were produced at a 5km x 5km resolution for each
751 calendar month and then aggregated to the level of the five Brazilian regions used in
752 this study (**Extended Data Fig. 6**). For consistency, we rescaled monthly suitability
753 values so that the sum of all monthly maps equalled the annual mean map⁹.

754 We then assessed the correlation between monthly *Aedes aegypti* climatic suitability
755 and the number of weekly ZIKV notified cases in each Brazilian region, to test how
756 well vector suitability explains the variation in the number of ZIKV notified cases. To
757 account for the correlation in each Brazilian region we fit a linear regression model
758 with a lag and two breakpoints. As there may be a lag between trends in suitability
759 and trends in notified cases, we include a temporal term in the model to allow for a
760 shift in the respective curves. Thus for each region, different sets of the constant and
761 linear terms are fitted to different time periods. More formally,

762

$$763 \quad \log(y_i + 1) = \alpha + \mathbb{I}(i \notin T)\alpha' + [b + \mathbb{I}(i \notin T)b']x_{i-l} \quad (4)$$

764

765 where y_i represents notified cases in a particular region in month i , x_i is the climatic
766 suitability in that region in month i , l is the time lag that yields the highest correlation
767 between y_i and x_i and T is the set of time indexes in the correlated region.

768 We then find the values of T and l that provide the highest adjusted- R^2 by stepwise
769 iterative optimisation. For each value of T evaluated, the optimal value of l (i.e. that
770 which gives the highest adjusted- R^2 for the model above) is found by the optim
771 function in R^{57} . Climatic suitability values were only calculated for each month, so to
772 calculate suitability values for any given point in time we interpolated between the
773 monthly values using a linear function. We found no significant effect of residual
774 autocorrelation in our data (**Extended Data Fig. 7**).

775

776

777

778 **Data availability**

779 Sequences of the primers and probes used here have been available at
780 <http://www.zibraproject.org> since the beginning of the project. XML files and datasets
781 analysed in this study are available from the same website. New Brazilian sequences
782 are available in GenBank under accession numbers KY558989 to KY559032 and
783 KY817930. New Colombian and Mexican sequences are available under accession
784 numbers KY317936-40 and KY606271-4, respectively. See **Extended Data Table 2**
785 for further details.

786

787

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871

872

873 **Extended Data Figure Legends**

874 **Extended Data Fig. 1. a.** The distribution of CT-values for the RT-qPCR+ samples
875 tested during the ZiBRA journey in Brazil ($n=181$ samples; median CT = 35.96). **b.**
876 shows the distribution of the temporal lag between the date of onset of clinical
877 symptoms and the date of sample collection of RT-qPCR+ samples (median lag = 2
878 days). Red dashed lines represent the median of the distributions. **(c)** Validation of
879 sequencing approaches. A phylogeny of the ZIKV Asian genotype estimated using
880 PhyML³⁷ is shown. The expanded clade highlighted in blue contains the WHO
881 reference ZIKV sequence¹⁹ (accession number KX369547), which was generated
882 using Illumina MiSeq. Sequences generated using MinION chemistries R9.4 2D, R9.4
883 1D, R9 1D, R9 2D and R7.3 2D contain no nucleotide differences and hence were
884 also placed in this clade. Scale bars represent expected nucleotide substitutions per
885 site (s/s). Am-ZIKV=American Zika virus lineage.

886
887 **Extended Data Fig. 2.** Temporal signal of the ZIKV Asian genotype. The correlation
888 between sampling dates and genetic distances from the tips to the root of a maximum
889 likelihood (ML) tree, estimated using PhyML³⁷, was explored using TempEst⁴⁵. **a.**
890 Estimates for the dataset used in the phylogenetic analysis presented in **Fig. 3c**, and **b.**
891 estimates for the same dataset with the addition of the P6-740 strain sampled in 1966
892 (accession number HQ234499).

893
894 **Extended Data Fig. 3.** A non-clock maximum likelihood phylogeny of our ZIKV
895 data set. Bootstrap branch support values are shown at each node. The phylogeny was
896 estimated using PhyML³⁷. Sequences generated in this study are highlighted in red.
897 Scale bar represents expected nucleotide substitutions per site.

898
899 **Extended Data Fig. 4.** Ancestral node location posterior probabilities (ANLPP), for
900 nodes A, B and C, estimated using the complete dataset (top row) and ten replicate
901 subsampled data sets (other rows). See **Methods** for details. ANLPPs were calculated
902 using two approaches: DTA=discrete trait analysis method³⁰ (left side columns) and
903 BASTA=Bayesian structured coalescent approximation method²⁹ (right side
904 columns). For each method, we employed an asymmetric model of location exchange
905 to estimate ancestral node locations and to infer patterns of virus spread among
906 regions.

907
908 **Extended Data Fig. 5.** Epidemic growth rates estimated from weekly ZIKV notified
909 cases in Brazil. Time series show the number of ZIKV notified cases in each region of
910 Brazil. Periods from which exponential growth were estimated are highlighted in
911 grey.

912
913 **Extended Data Fig. 6.** Seasonal suitability for ZIKV transmission in the Americas.
914 These maps were estimated by collating data on *Aedes* mosquitoes, temperature,
915 relative humidity and precipitation, and are the basis of the trends in suitability for
916 different regions shown in main text **Figs. 1 and 4**. For method details, see ^{9,60}.

917
918 **Extended Data Fig. 7.** Partial autocorrelation functions for the linear model
919 associating climatic suitability and ZIKV notified cases in each geographic region in
920 Brazil. The residuals for the North, Northeast, Centre-West and Southeast regions
921 show no autocorrelation, while a small amount of autocorrelation cannot be excluded
922 for the South region.

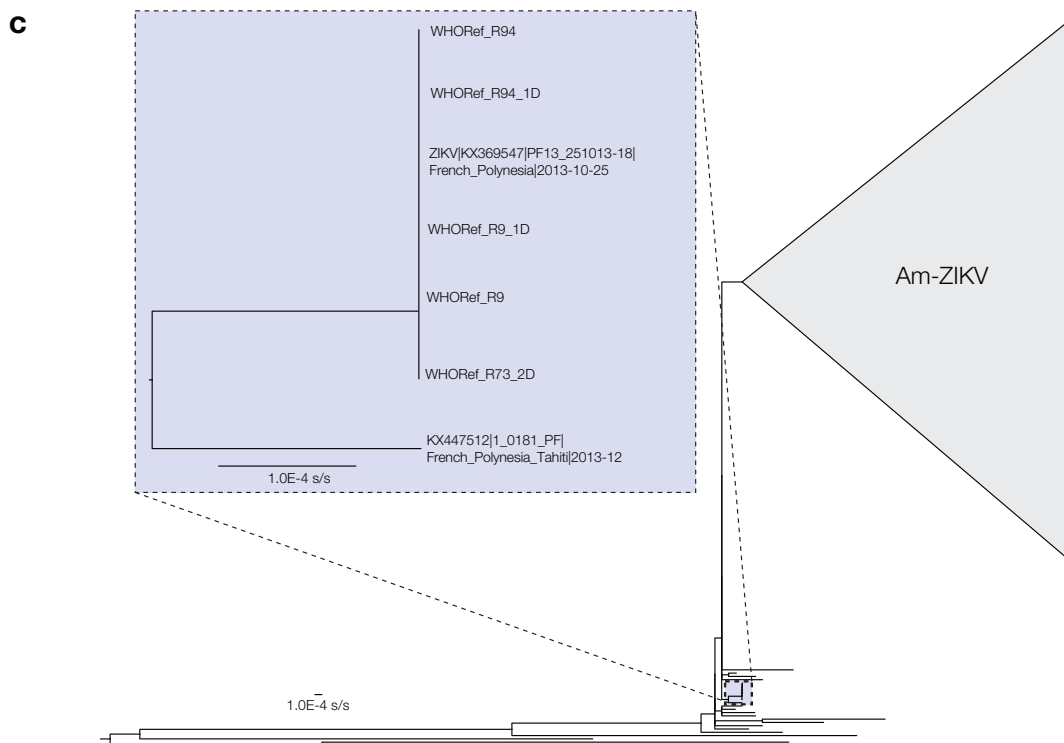
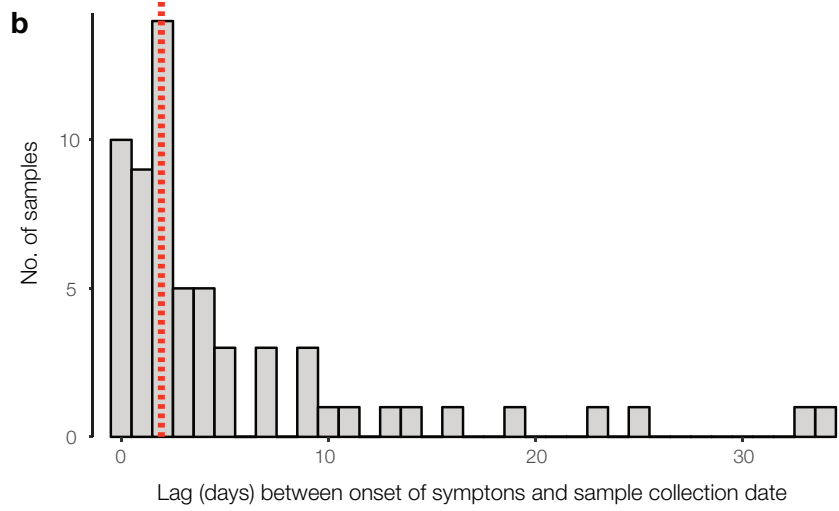
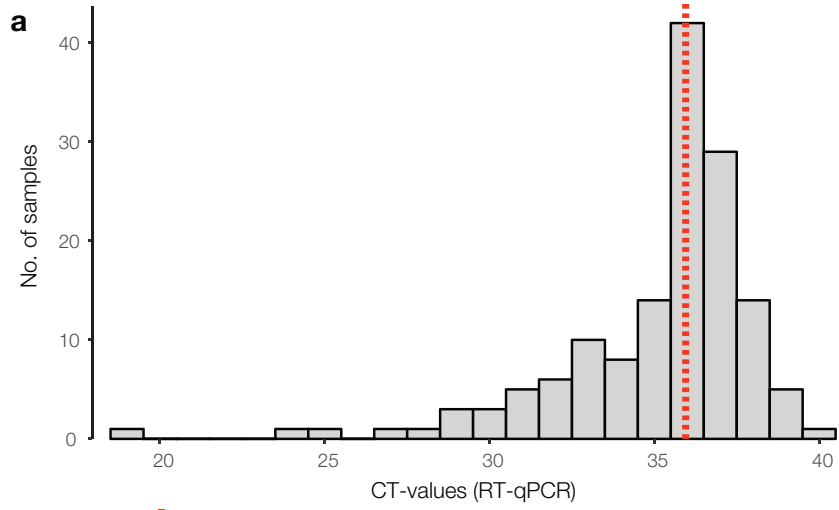
923 **Extended Data Table Legends**

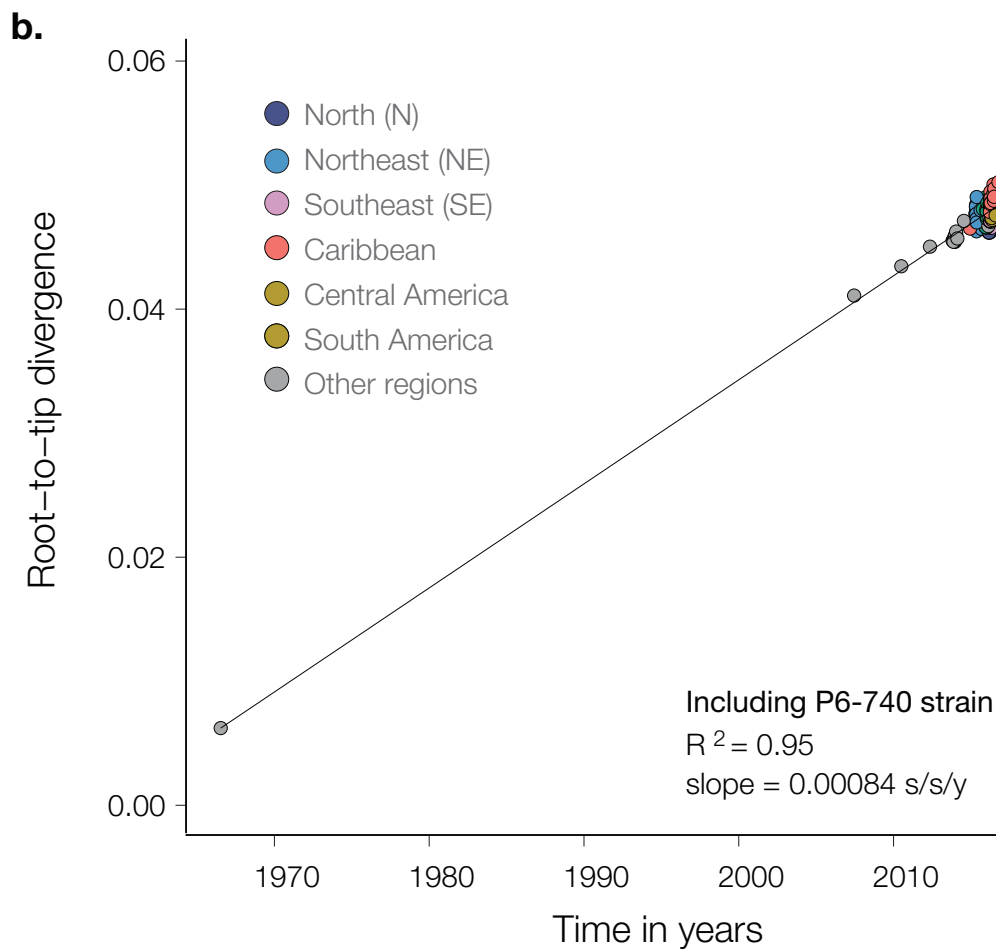
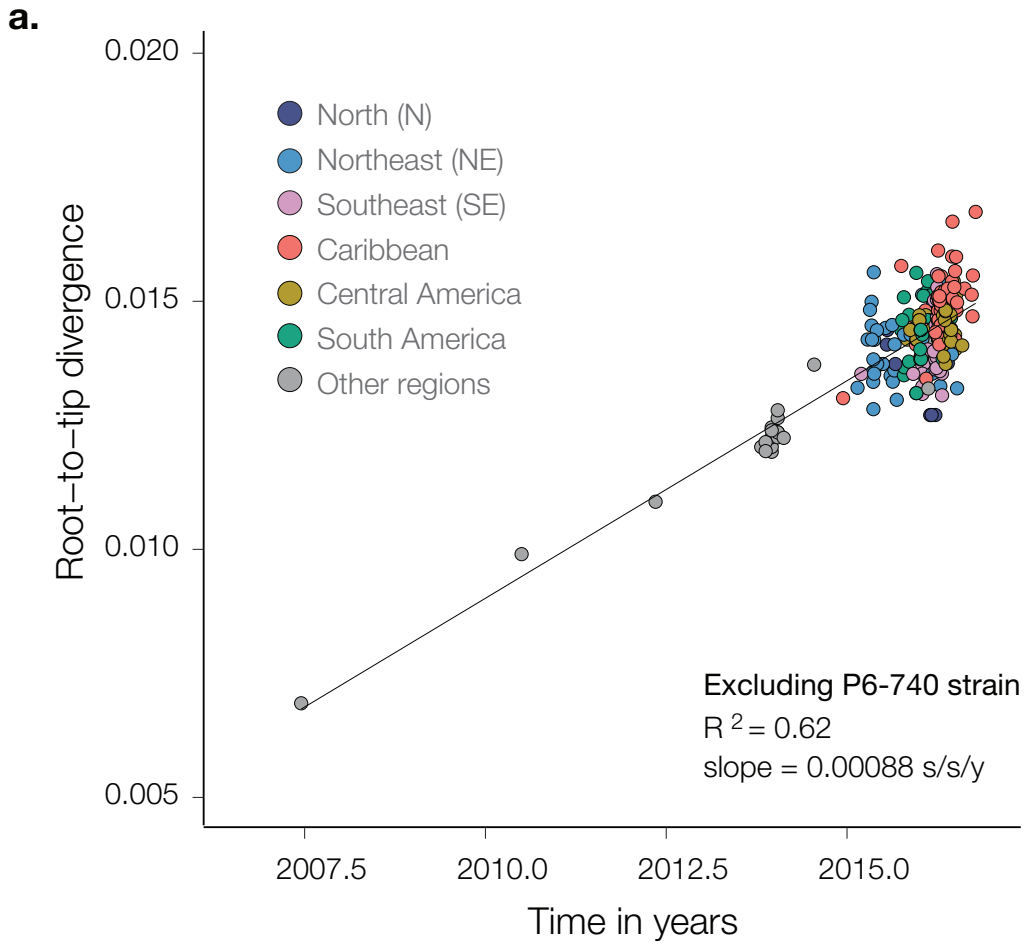
924 **Extended Data Table 1. a.** Summary of the clinical samples tested ($n=1330$, of
925 which 181 were RT-qPCR+) by the ZiBRA mobile lab in June 2016, NE Brazil. 84%
926 of samples with known collection dates ($n=698$ of 826) were from 2016. ZIKV
927 notified cases were confirmed using RT-qPCR (see **Methods**). Collection lag
928 represents the median time interval (in days) between the date of onset of clinical
929 symptoms and date of sample collection (both dates available for $n=219$) for all
930 samples (including those that subsequently tested RT-qPCR negative). Federal states
931 are RN: Rio Grande do Norte, PB: Paraíba, PE: Pernambuco, AL: Alagoas, BA:
932 Bahia. Sample numbers in the FioCruz, PE row include RT-PCR+ cases from
933 Pernambuco generated at Fiocruz Pernambuco. **b.** Parameters of the model
934 measuring the link between climatic vector suitability and notified ZIKV cases in
935 different Brazilian regions (CW: Centre-West, N: North, NE: Northeast, SE:
936 Southeast, S: South). For each region, the table provides the estimated correlated
937 time period (T), P-value of the linear term of suitability in T , adjusted- R^2 of the
938 model, and time lag (l). **c.** For each region, estimates of the basic reproductive
939 number (R) of ZIKV are shown for several values of generation time (g) parameter,
940 together with the corresponding estimates of exponential growth rate (r) (per day)
941 obtained from notified ZIKV case counts (see Extended Data Fig. 7). 1st: epidemic
942 wave in 2015; 2nd: epidemic wave in 2016.

943
944 **Extended Data Table 2.** Sequencing statistics. Accession numbers, sample IDs,
945 sequencing coverage, RT-qPCR values and epidemiological information for the
946 samples from Brazil generated in this study. For the sequences from RJ state,
947 alignments were performed against version 2 (KJ776791.2) of the genome reference;
948 all other sequences used version 1 (KJ776791.1).

949
950 **Extended Data Table 3. a.** Estimated per-gene rates of evolution (mean and 95%
951 Bayesian credible intervals=BCIs) are shown in units of 10^{-3} substitutions per site per
952 year. **b.** Log-marginal likelihood estimates using the path-sampling (PS) and
953 Stepping-Stone (SS) model selection approaches⁴⁷. The overall ranking of the models
954 is shown in parentheses for each estimator and the best-fitting combination is
955 underscored. Two molecular clock models were tested here. SC: Strict clock model,
956 UCLN: uncorrelated relaxed clock with lognormal distribution⁴⁶. **c.** Estimated dates
957 of nodes A, B and C (**Fig. 3**) under various different molecular clock and coalescent
958 model combinations. TMRCA: time of the most recent common ancestor, BCI:
959 Bayesian credible interval, SC: strict molecular clock model, UCLN: uncorrelated
960 clock with lognormal distribution.

961





Ancestral node location posterior probability (DTA)

Ancestral node location posterior probability (BASTA)



ANLPP

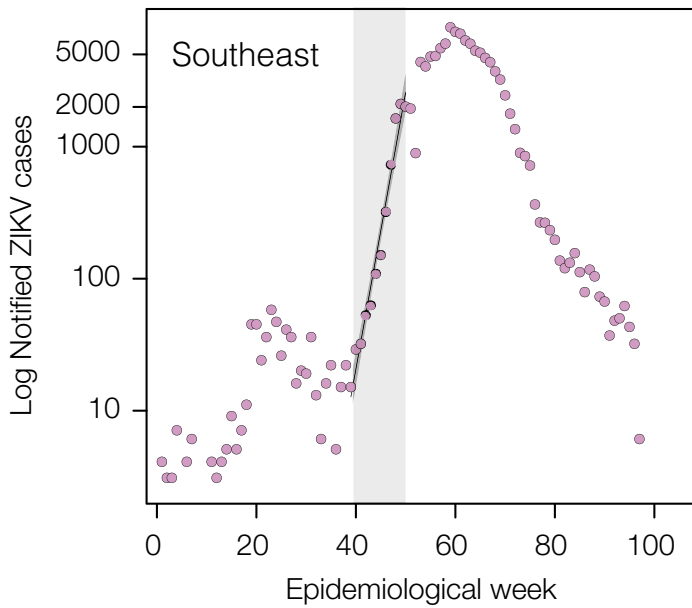
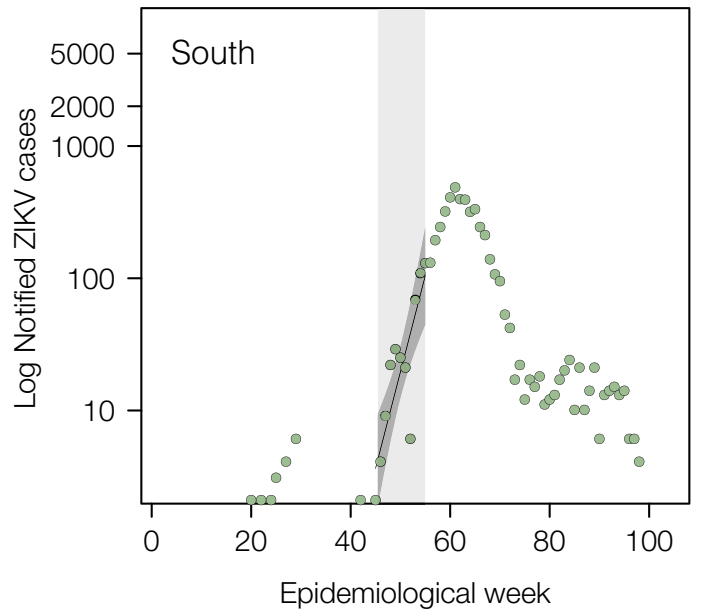
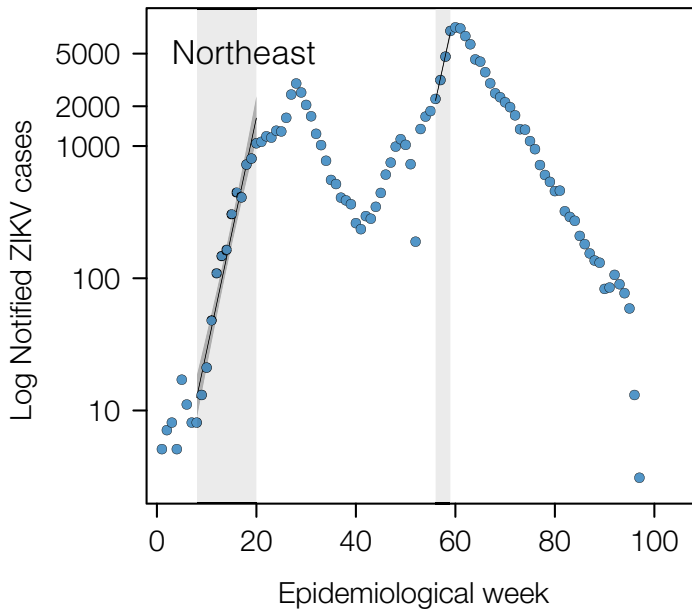
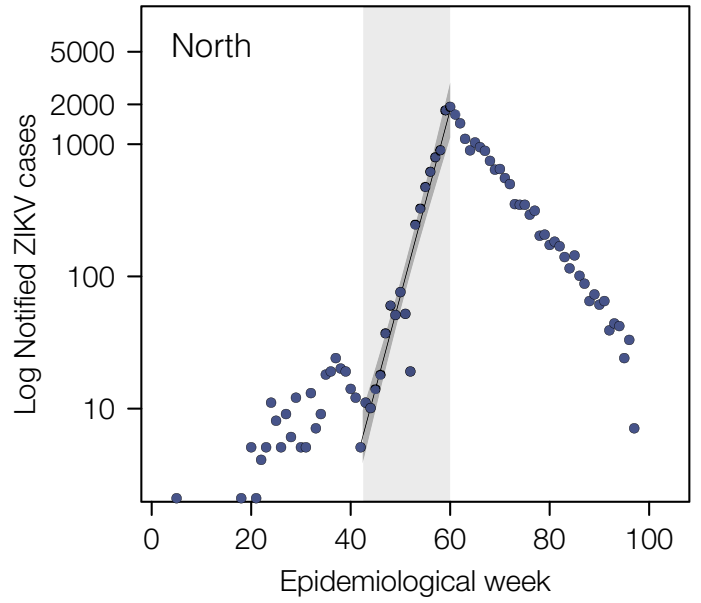
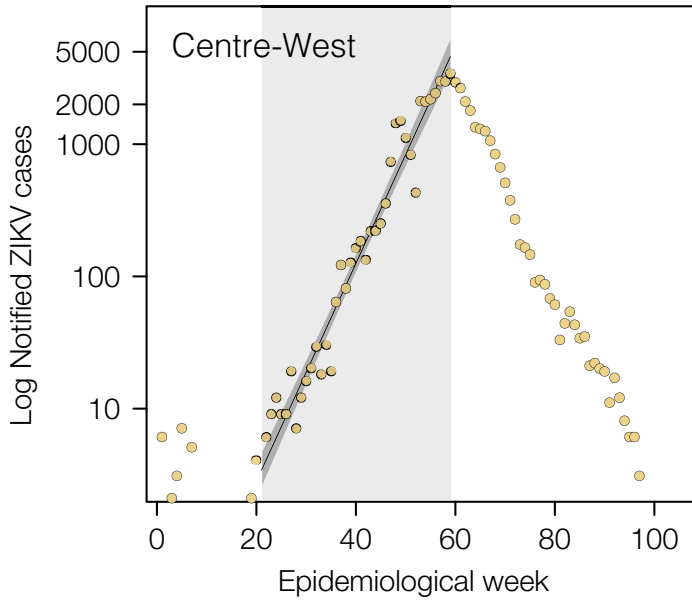
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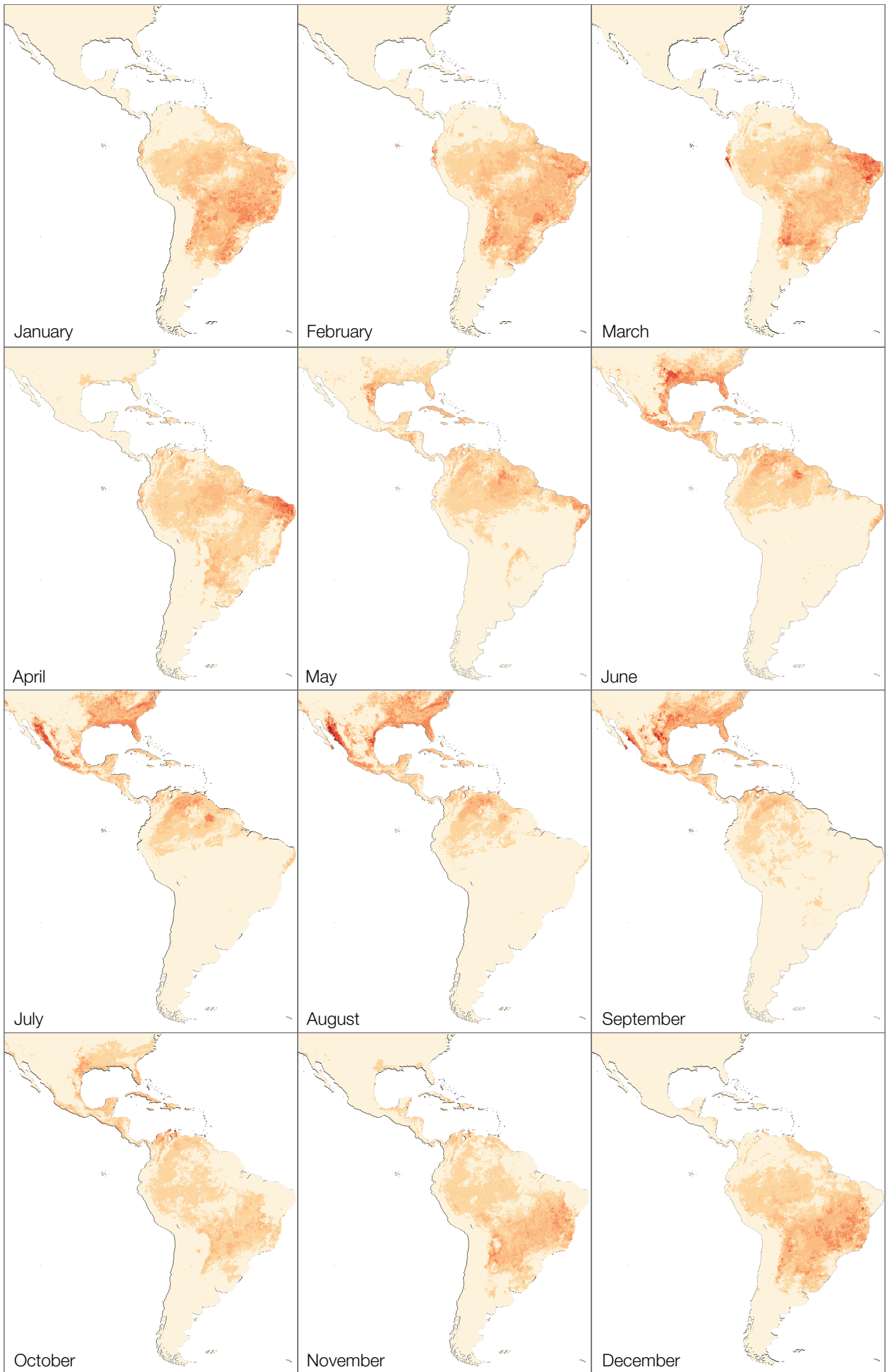
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ANLPP

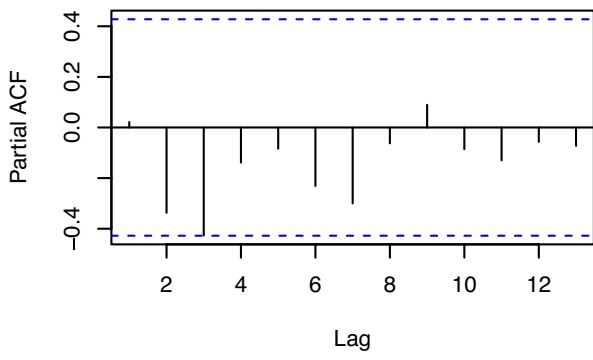
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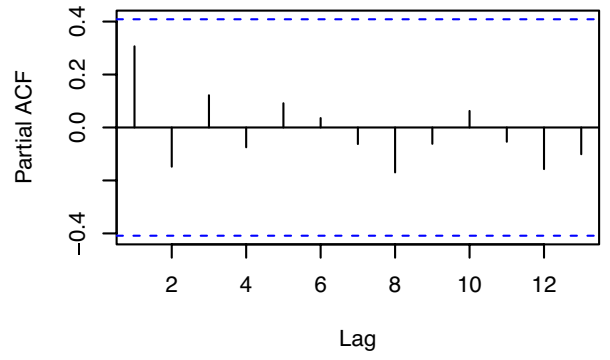




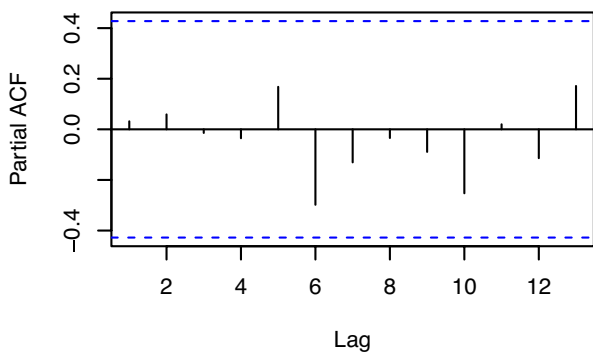
North



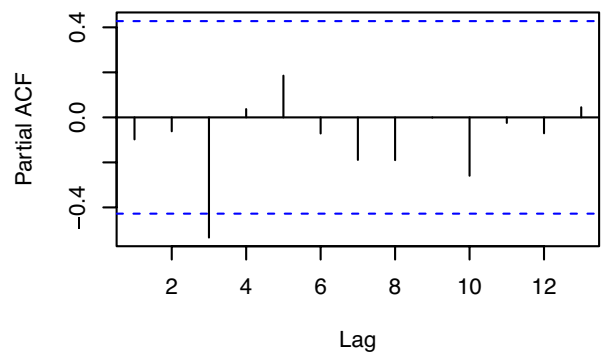
Northeast



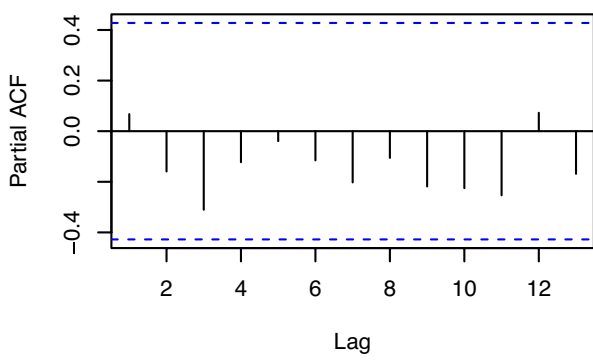
Centre-West



South



Southeast



(a)

Laboratory, Federal state	No. Positives / Tested (%)	Ct value (mean, min- max)	Collection lag (median, min-max)
LACEN, RN	27/335 (8.1%)	35.9 (18.6-39.1)	5 (4-16)
LACEN, PB	26/276 (9.4%)	35.7 (30.7-37.0)	6 (0-88)
FioCruz, PE	95/315 (30%)	34.6 (24.1-38.3)	2.5 (0-33)
LACEN, AL	16/140 (11%)	34.1 (27.1-40.2)	2 (0-3)
FioCruz, BA	17/264 (6.4%)	35.8 (24.7-39.2)	4 (0-228)

(b)

	N	NE	CW	S	SE
Correlated time period	12/2015 to 10/2016	7/2015 to 10/2016	9/2015 to 8/2016	6/2015 to 05/2016	11/2015 to 9/2016
<i>P</i> -value	<0.0001	0.00013	<0.0001	<0.0001	<0.0001
Adjusted- <i>R</i> ²	0.929	0.8448	0.987	0.9543	0.953
Time lag (months)	1.27	0	1.12	1.19	1.33

(c)

Region	<i>R</i> (mean, CI), <i>g</i> =20 days	<i>R</i> (mean, CI), <i>g</i> =15 days	<i>R</i> (mean, CI), <i>g</i> =10 days	Growth rate (<i>r</i> , CI)
CW	1.71 (1.65-1.78)	1.46 (1.20-1.77)	1.29 (1.13-1.46)	0.027 (0.02-0.03)
N	2.48 (2.19-2.81)	1.98 (1.80-2.18)	1.58 (1.48-1.69)	0.046 (0.04-0.05)
NE, 1 st	3.12 (2.69-3.60)	2.36 (2.11-2.63)	1.78 (1.65-1.91)	0.06 (0.05-0.07)
NE, 2 nd	3.03 (2.74-3.36)	2.31 (2.14-2.49)	1.75 (1.66-1.84)	0.06 (0.05-0.06)
SE	3.85 (3.35-4.42)	2.77 (2.49-3.07)	1.98 (1.84-2.12)	0.07 (0.06-0.076)
S	2.57 (1.72-3.82)	2.04 (1.50-2.75)	1.61 (1.31-1.97)	0.05 (0.04-0.07)

Accession Number	Sample ID	Aligned Reads	Consensus nucleotide bases (% of reference)	RT-qPCR Ct	Collection Date	Municipality	State
KY558989	ZBRA105	58128	9846 (92)	29.5	2015-02-23	João Câmara	RN
KY558990	ZBRC14	19111	8612 (81)	32.81	2016-01-15	Recife	PE
KY558991	ZBRC16	9161	7178 (67)	34.94	2016-01-19	Garanhuns	PE
KY558992	ZBRC18	7183	7459 (70)	35.14	2016-01-06	Caetes	PE
KY558993	ZBRC25	20533	5688 (53)	35.89	2016-01-18	Sanhoro	PE
KY558994	ZBRC28	7905	8987 (84)	36.02	2016-01-18	Limoeiro	PE
KY558995	ZBRC301	20826	9843 (92)	31.99	2015-05-13	Paulista	PE
KY558996	ZBRC302	26331	10007 (94)	30.78	2015-05-13	Paulista	PE
KY558997	ZBRC303	12575	5873 (55)	32.81	2015-05-14	Olinda	PE
KY558998	ZBRC313	16530	9478 (89)	30.77	2015-06-15	Paulista	PE
KY558999	ZBRC319	17316	10565 (99)	24.07	2016-07-10	Olinda	PE
KY559000	ZBRC321	11434	8647 (81)	30.62	2015-08-09	Paulista	PE
KY559001	ZBRD103	13192	8380 (78)	29.09	2015-08-20	Murici	AL
KY559002	ZBRD107	77118	7415 (69)	30.31	2015-09-09	Maceió	AL
KY559003	ZBRD116	21211	9785 (92)	27.13	2015-08-28	Arapiraca	AL
KY559004	ZBRE69	2313	6866 (64)	24.72	2016-04-16	Feira de Santana	BA
KY559005	ZBRX1	21267	10559 (99)	25	2016-04-18	Ribeirão Preto	SP
KY559006	ZBRX2	24105	9961 (93)	32	2016-04-18	Ribeirão Preto	SP
KY559007	ZBRX4	14722	10563 (99)	26	2016-04-18	Ribeirão Preto	SP
KY559008	ZBRX6	12516	6893 (64)	33	2016-04-19	Ribeirão Preto	SP
KY559009	ZBRX7	10981	8563 (80)	33	2016-04-19	Ribeirão Preto	SP
KY559010	ZBRX8	7445	8702 (81)	33	2016-04-19	Ribeirão Preto	SP
KY559011	ZBRX11	21214	9379 (88)	31	2016-04-19	Ribeirão Preto	SP
KY559012	ZBRX12	19838	10305 (97)	31	2016-04-19	Ribeirão Preto	SP
KY559013	ZBRX13	11809	10564 (99)	21	2016-04-24	Ribeirão Preto	SP
KY559014	ZBRX14	5873	7469 (70)	33	2016-04-24	Ribeirão Preto	SP
KY559015	ZBRX15	20190	10563 (99)	27	2016-04-24	Ribeirão Preto	SP
KY559016	ZBRX16	9698	9027 (85)	32	2016-04-25	Ribeirão Preto	SP
KY559017	ZBRX100	5976	9609 (90)	28.5	2016-05-19	Ribeirão Preto	SP
KY559018	ZBRX102	13990	9508 (89)	33.91	2016-02-25	Porto Nacional	TO
KY559019	ZBRX103	17635	9514 (89)	36.76	2016-05-24	Araguaina	TO
KY559020	ZBRX106	29877	8458 (79)	32.36	2016-03-07	Palmas	TO
KY559021	ZBRX127	18914	10066 (94)	29.6	2016-03-10	Palmas	TO
KY559022	ZBRX128	18480	8650 (81)	28.79	2016-03-13	Palmas	TO
KY559023	ZBRX130	16667	9914 (93)	29.06	2016-03-22	Palmas	TO
KY559024	ZBRX137	15895	9767 (91)	34.83	2016-03-03	Palmas	TO
KY559025	ZBRY1	41036	8941 (84) †	33.53	2016-01	Rio de Janeiro	RJ
KY559026	ZBRY4	27865	8433 (79) †	34.21	2016-01	Rio de Janeiro	RJ
KY559027	ZBRY6	11779	10300 (97) †	22.66	2016-01	Rio de Janeiro	RJ
KY559028	ZBRY12	4980	3061 (28) †	33.66	2016-01	Rio de Janeiro	RJ
KY559029	ZBRY11	18530	5873 (55) †	31.11	2016-01	Rio de Janeiro	RJ
KY559030	ZBRY10	14067	5712 (53) †	30.84	2016-01	Rio de Janeiro	RJ
KY559031	ZBRY8	5708	9184 (86) †	30.96	2016-01	Rio de Janeiro	RJ
KY559032	ZBRY7	7749	9018 (84) †	28.07	2016-01	Rio de Janeiro	RJ
KY817930	ZBRY14	8040	5389 (50) †	34.2	2016-02-15	Rio de Janeiro	RJ

(a)

Gene	Mean	Lower BCI	Upper BCI
<i>C</i>	0.86	0.65	1.06
<i>prM</i>	0.98	0.85	1.12
<i>E</i>	1.04	0.87	1.24
<i>NS1</i>	0.97	0.83	1.12
<i>NS2A</i>	0.98	0.83	1.13
<i>NS2B</i>	1.12	0.93	1.34
<i>NS3</i>	0.93	0.75	1.11
<i>NS4A</i>	0.87	0.74	1.01
<i>NS4B</i>	1.11	0.9	1.35
<i>NS5</i>	1.35	0.87	1.12

(b)

Clock	Coalescent	PS	SS
UCLN	Skyline	-32090.664	-32116.195
SC	Skyline	-32117.581	-32148.760
UCLN	Exponential	-32193.426	-32218.348
UCLN	Constant	-32206.219	-32234.196
SC	Constant	-32229.262	-32257.900
SC	Exponential	-32244.500	-32270.815

(c)

Clock model	Coalescent prior	Node A TMRCA (95% BCIs)	Node B TMRCA (95% BCIs)	Node C TMRCA (95% BCIs)
SC	Constant	2013.59 (2013.4,2013.77)	2013.83 (2013.6,2014.05)	2013.90 (2013.65,2014.12)
SC	Exponential	2013.59 (2013.38,2013.77)	2013.82 (2013.58,2014.04)	2013.89 (2013.65,2014.11)
SC	Skyline	2013.66 (2013.48,2013.81)	2013.93 (2013.74,2014.14)	2013.99 (2013.75,2014.18)
UCLN	Constant	2013.65 (2013.42,2013.84)	2013.91 (2013.63,2014.2)	2014.04 (2013.73,2014.32)
UCLN	Exponential	2013.66 (2013.45,2013.84)	2013.88 (2013.64,2014.13)	2014 (2013.73,2014.25)
UCLN	Skyline	2013.71 (2013.54,2013.85)	2014.03 (2013.76,2014.26)	2014.16 (2013.89,2014.41)