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Genomic epidemiology reveals multiple introductions of 1 Zika virus into the United States 2

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Zika virus (ZIKV) is causing an unprecedented epidemic linked to severe congenital syndromes^{1,2}. 56 57 In July 2016, mosquito-borne ZIKV transmission was reported in the continental United States and since then, hundreds of locally-acquired infections have been reported in Florida^{3,4}. To gain insights 58 59 into the timing, source, and likely route(s) of ZIKV introduction, we tracked the virus from its first 60 detection in Florida by sequencing ZIKV genomes from infected patients and Aedes aegypti mosquitoes. We show that at least four introductions, but potentially as many as 40, contributed to 61 62 the outbreak in Florida and that local transmission likely started in the spring of 2016 - several 63 months before initial detection. By analyzing surveillance and genetic data, we discovered that 64 ZIKV moved among transmission zones in Miami. Our analyses show that most introductions are 65 linked to the Caribbean, a finding corroborated by the high incidence rates and traffic volumes 66 from the region into the Miami area. Our study provides an understanding of how ZIKV initiates 67 transmission in new regions.

ZIKV transmission in the Americas was first reported in Brazil in May 2015⁵, though the virus was likely 68 introduced 1-2 years prior to its detection^{6–8}. By January 2016, ZIKV cases were reported from several South and Central American countries and most islands in the Caribbean⁹. Like dengue virus (DENV) and chikungunya virus (CHIKV), ZIKV is vectored primarily by *Aedes* mosquitoes^{10–13}. The establishment of the peridomestic species *Ae. aegypti* in the Americas¹⁴ has facilitated DENV, CHIKV, and now likely 69 70 71 72 ZIKV to become endemic in this region¹⁵. In the continental United States, transient outbreaks of DENV and CHIKV have been reported in regions of Texas and Florida^{4,16–21} with abundant seasonal *Ae. aegypti* 73 74 populations^{14,22}. 75

76 The 2016 ZIKV outbreak in Florida generated 256 confirmed ZIKV infections⁴ (Fig. 1a). While 77 transmission was confirmed across four counties in Florida (Fig. 1b), the outbreak was most intense in 78 Miami-Dade County (241 infections). Although the case location could not always be determined, at least 79 114 (47%) infections were likely acquired in one of three distinct transmission zones: Wynwood, Miami

80 Beach, and Little River (Fig. 1c-d).

81 Using mosquito surveillance data, we determined the extent of mosquito-borne ZIKV transmission in 82 Miami. Of the 24,351 mosquitoes collected from June to November 2016, 99.8% were Ae, aegypti and 8 83 pools of \leq 50 mosquitoes tested positive for ZIKV (Fig. 1c, Extended Data Fig. 1). From these pools, we 84 estimated that ~1 out of 1,600 Ae. aegypti mosquitoes were infected (0.061%, 95% CI: 0.028-0.115%, Extended Data Fig. 1a). This is similar to infection rates during DENV and CHIKV outbreaks²³. Although 85 86 we did not detect ZIKV-infected mosquitoes outside Miami Beach (Fig. 1c), we found that the number of 87 human ZIKV cases correlated strongly with Ae. aegypti abundance within each transmission zone 88 (Spearman r = 0.61, Fig. 1d, Extended Data Fig. 1b). This suggests that *Ae. aegypti* mosquitoes were the 89 primary mode of transmission and that changes to vector abundance impacted human infection rates. We found that the application of insecticides³ suppressed mosquito populations during periods of intensive 90 91 usage (Extended Data Fig. 1c), and therefore likely contributed to ZIKV clearance.

92 We sequenced 39 ZIKV genomes from clinical and mosquito samples without cell culture²⁴ 93 (Supplementary Table 1a). Our ZIKV dataset included 29 genomes from patients with locally-acquired 94 infections (Fig. 1d) and 7 from Ae. aegypti pools (Fig. 1c). We also sequenced 3 ZIKV genomes from 95 travel-associated cases from Florida. Our dataset included cases from all transmission zones in Miami 96 (Fig. 1d) and represented ~11% of all confirmed locally-acquired cases in Florida. We made all sequence 97 data openly available (PRJNA342539, PRJNA356429) immediately after data generation.

98 We reconstructed phylogenetic trees from our ZIKV genomes along with 65 published genomes from 99 other affected regions (Fig. 2, Extended Data Fig. 2 and 3). We found that the Florida ZIKV genomes 100 formed four distinct lineages (labeled F1-F4, Fig. 2a), three of which (F1-F3) belonged to the same clade

101 (labeled A, Fig. 2a). We only sampled a single human case each from the F3 and F4 lineages, consistent with limited transmission (Fig. 2a). The other two Florida lineages (F1-F2) comprised ZIKV genomes
 from human and mosquito samples within Miami-Dade County (Fig. 2b).

Using time-structured phylogenies²⁵, we estimated that at least four separate introductions were 104 responsible for the locally-acquired cases observed in our dataset. The phylogenetic placement of lineage 105 106 F4 clearly indicates that it resulted from an independent introduction of a lineage distinct from those in 107 clade A (Fig. 2a). For the two well-supported nodes linking lineages F1-F2 (labeled B, Fig. 2a) and F1-F3 108 (A, Fig. 2a), we estimated the time of the most recent common ancestor (tMRCA) to be during the 109 summer of 2015 (95% highest posterior density [HPD]: June-September, 2015). Our data displayed a strong clock signal (Extended Data Fig. 2b) and tMRCA estimates were robust across a range of models 110 (Extended Data Table 1a). Thus while F1-F3 belong to clade A, any fewer than three distinct 111 112 introductions leading to these lineages would have required undetected transmission of ZIKV in Florida 113 for approximately one year (Fig. 2a).

To estimate the likelihood of a single ZIKV transmission chain persisting for over a year, we modeled spread under different assumptions of the basic reproductive number (R_0). Using the number of locallyacquired and travel-associated cases, along with the number of observed genetic lineages, we estimated an R_0 between 0.5 and 0.8 in Miami-Dade County (Extended Data Fig. 4). Even at the upper end of this range, the probability of a single transmission chain persisting for over a year is extremely low (~0.5%, Fig. 2c). This is especially true considering the low *Ae. aegypti* abundance during the winter months (Extended Data Fig. 1d).

121 Given the low probability of long-term persistence, we expect that our ZIKV genomes (F1-F4) were the 122 result of at least four introductions. Differences in surveillance practices and a high number of travel-123 associated cases (Fig. 1a), however, likely mean that unsampled ZIKV introductions also contributed to 124 the outbreak. To estimate the total number of ZIKV introductions, we modeled scenarios that resulted in 125 241 locally-acquired cases within Miami-Dade County, and found that with R_0 values of 0.5-0.8, we expect 17-42 (95% CI 3-63) separate introductions to have contributed to the outbreak (Fig. 2d). The 126 127 majority of these introductions would likely have generated a single secondary case that was undetected 128 in our genetic sampling (Extended Data Fig. 4a). Incorporating under-reporting in a sensitivity analysis 129 increases R_0 estimates slightly to 0.7-0.9 (Extended Data Fig. 4f-i).

130 The two main ZIKV lineages, F1 and F2, included the majority of genomes from Florida (92%, Fig. 2a). Assuming they represent two independent introductions, we estimated when each of these lineages 131 132 arrived in Florida. The probability densities for the tMRCAs of both F1 and F2 were centered around 133 March-April, 2016 (Fig. 2b, 95% HPD: January-May, 2016). The estimated timing for these introductions corresponds with suitable Ae. aegypti populations in Miami-Dade County²⁶ (Extended Data Fig. 1d) and 134 suggests that ZIKV transmission could have started at least two months prior to its detection in July 2016 135 136 (Fig. 1a). The dates of the introductions could be more recent if multiple F1 or F2 lineage viruses arrived 137 independently. However, more than 2 introductions would be necessary to substantially change our 138 estimates for the timing of the earliest introduction.

139 To understand transmission dynamics within Miami, we analyzed our genomic data together with case 140 data from the Florida Department of Health (DOH, Supplementary Table 1a). While spatially distinct, the 141 three ZIKV transmission zones occurred within \sim 5 km of each other (Fig. 1c) and we found that the 142 ZIKV infections associated with each zone overlapped temporally (Fig. 1d). Our ZIKV genomes with 143 zone assignments all belonged to lineages F1 and F2, but neither of these lineages were confined to a 144 single zone (Fig. 2b). In fact, we detected both F1 and F2 lineage viruses from Ae. aegypti collected from 145 the same trap 26 days apart (mosquitoes 5 and 8, Fig. 2b). These findings suggest that ZIKV moved 146 among areas of Miami.

147 Determining the sources and routes of ZIKV introductions could help mitigate future outbreaks. We 148 found that lineages F1-F3 clustered with ZIKV genomes sequenced from the Dominican Republic and Guadeloupe (Fig. 2, Extended Data Fig. 2 and 3). In contrast, F4 clustered with genomes from Central 149 150 America (Fig. 2, Extended Data Fig. 2 and 3). These findings suggest that while ZIKV outbreaks occurred 151 throughout the Americas, the Caribbean islands were the main source of establishing local ZIKV 152 transmission in Florida. Because of severe undersampling of ZIKV genomes, however, we cannot rule out 153 other source areas. Similarly, even though we found that the Florida ZIKV genomes clustered together 154 with sequences from the Dominican Republic, our results do not prove that ZIKV entered Florida from 155 this country.

156 We investigated ZIKV infection rates and travel patterns to corroborate our phylogenetic evidence for 157 Caribbean introductions. We found that the Caribbean islands bore the highest ZIKV incidence rates (Fig. 158 2b), despite Brazil and Colombia reporting the highest absolute number of cases (January to June, 2016, 159 Fig. 3a, Extended Data Fig. 5, Supplementary Table 1b). During the same time period, we estimated that \sim 3 million travelers arrived from the Caribbean, accounting for 54% of the total traffic into Miami, with 160 the vast majority (~2.4 million) arriving via cruise ships (Fig. 3b, Extended Data Fig. 6, Supplementary 161 162 Table 1b). Combining the infection rates with travel capacities, we estimated that ~60-70% of ZIKV 163 infected travelers arrived from the Caribbean (Fig. 3c and Extended Data Fig. 7a). We also found that the 164 number of travel-associated ZIKV cases correlated strongly with the expected number of importations 165 from the Caribbean (Spearman r = 0.8, Fig. 3d, Extended Data Fig. 7b). Finally, 67% of the travel-166 associated infections in Florida reported recent travel to the Caribbean (Fig. 3e); however, their mode of 167 travel is unknown. Taken together, these findings suggest that a high incidence of ZIKV in the Caribbean, 168 combined with frequent travel, could have played a key role in the establishment of ZIKV transmission in 169 Florida. These findings, however, do not indicate that cruise ships themselves are risk factors for human 170 ZIKV infection, but only that they served as a major mode of transportation from areas with active 171 transmission. In addition, ZIKV exposure may vary among individuals depending on their purpose of 172 travel and therefore we cannot determine the specific contribution of ZIKV-infected travelers arriving via 173 airlines or cruise ships.

174 The majority of the Florida ZIKV outbreak occurred in Miami-Dade County (Fig. 1b). To determine if 175 there is a higher potential for ZIKV outbreaks in this area, we analyzed incoming passenger traffic from 176 regions with ZIKV transmission along with local Ae. aegypti abundance. We estimated that Miami and 177 nearby Fort Lauderdale received ~72% of traffic (Fig. 4) and Miami received more air and sea traffic 178 from ZIKV endemic areas than any other city in the United States (Extended Data Fig. 8). During January 179 to April 2016, we estimated that Ae. aegypti abundance was highest in southern Florida²² (Fig. 4, Extended Data Fig. 1d, Extended Data Fig. 8). By June, most of Florida and several cities across the South likely supported high *Ae. aegypti* populations^{14,22} (Extended Data Fig. 8); however, most of this 180 181 region has not reported local Ae. aegypti-borne virus transmission in at least 60 years¹⁹. In fact, the only 182 region outside of Florida with local ZIKV transmission is southern Texas²⁷, which is also the only other region with recent DENV outbreaks^{19–21}. Therefore, the combination of travelers, mosquito ecology, and 183 184 185 human population density likely make Miami one of the few places in the continental United States at risk for Ae. aegypti-borne virus outbreaks^{22,26,28}. 186

The extent of ZIKV transmission in Florida was unprecedented, with more reported ZIKV cases in 2016 (256) than DENV cases since 2009 (136)^{4,16,17}. This case difference may be reflected by lower incidence of endemic DENV than epidemic ZIKV in source countries^{29,30}, resulting in fewer DENV importations (reported travel cases since 2009: 654 DENV and 1,016 ZIKV)⁴. Given that the majority of ZIKV infections are asymptomatic^{2,31}, the true number of ZIKV cases was likely much higher. Despite this, we estimated that the average R_0 was less than 1 and therefore multiple introductions were necessary to give rise to the observed outbreak³². The high volume of traffic entering Florida from ZIKV-affected regions, especially the Caribbean, likely provided a substantial supply of ZIKV-infected individuals³³. Because

- Florida is unlikely to sustain long-term ZIKV transmission³², the potential for future ZIKV outbreaks in this region is dependent upon activity elsewhere. Therefore, we expect that outbreaks in Florida will cycle with the ZIKV transmission dynamics in the Americas^{7,8,15}.

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220 Author Contributions

221 All contributions are listed in order of authorship. Designed the experiments: N.D.G., J.T.L., G.D., 222 M.U.G.K., D.A.T.C., P.C.S, L.D.G., S.F.M., T.B., O.G.P., S.I., G.P., and K.G.A.; Collected samples: 223 A.L.T., S.W., D.M.M., A.B., L.M.P., D.P., P.N.L., M.R., V.K.B., D.I.W., M.R.C., E.W.K., K.N.H., 224 A.C.C., R.J., M.C.P., C.V., D.S., L.D.G., S.F.M., and S.I.; Performed the sequencing: N.D.G., M.W.R., 225 K.P., D.R., R.R.-S., G.O., and E.N.; Provided data, reagents, or protocols: N.D.G., J.T.L., G.D., 226 M.U.G.K., K.G., M.R.W., R.R.-S., G.O., H.C.M., M.L.B., K.G.B., B.C., C.A.F., A.G.-Y., A.G., C.L., 227 B.M., C.B.M., D.J.P., J.Q., S.F.S., C.T.-T., K.L.M., S.M.W., S.W., N.L.Y., J.Q., J.R.F., K.K., S.E.B., 228 A.J.M., R.F.G., N.J.L., M.C.P., C.V., P.C.S., S.F.M., and S.I.; Analyzed the data: N.D.G., J.T.L., G.D., 229 M.U.G.K., K.G., J.T., J.R.F., R.C.R., N.R.F., D.A.T.C., A.K., M.S.-L., T.B., S.F.M, O.G.P., S.I., and 230 K.G.A.; Edited manuscript: G.D., M.U.G.K., J.T., S.F.S., A.R., T.B., O.G.P., S.I., and G.P.; Wrote 231 manuscript: N.D.G., J.T.L., and K.G.A.; All authors read and approved the manuscript.

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354

236 Figure Legends

237 Figure 1 | Zika virus outbreak in Florida. (a) Weekly counts of confirmed travel-associated and 238 locally-acquired ZIKV cases in 2016. (b) Four counties reported locally-acquired ZIKV cases in 2016: 239 Miami-Dade (241), Broward (5), Palm Beach (8), Pinellas (1), and unknown origin (1). (c) The locations 240 of mosquito traps and collected Ae. aegypti mosquitoes found to contain ZIKV RNA (ZIKV+) in relation to the transmission zones within Miami. (d) Temporal distribution of weekly ZIKV cases (left y-axis), 241 242 sequenced cases (bottom), and Ae. aegypti abundance per trap night (right y-axis) associated with the 243 three described transmission zones. ZIKV cases and sequences are plotted in relation to symptom onset 244 dates (n=18). Sequenced cases without onset dates or that occurred outside of the transmission zones are 245 not shown (n=10). Human cases and Ae. aegypti abundance per week were positively correlated 246 (Spearman r = 0.61, Extended Data Fig. 1b). The maps were generated using open source basemaps³⁴.

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248 Figure 2 | Multiple introductions of Zika virus into Florida. (a) Maximum clade credibility (MCC) 249 tree of ZIKV genomes sequenced from outbreaks in the Pacific islands and the epidemic in the Americas. Tips are colored based on collection location. The five tips outlined in blue but filled with a different 250 251 color indicate ZIKV cases in the United States associated with travel (fill color indicates the probable 252 location of infection). Clade posterior probabilities are indicated by white circles filled with black relative 253 to the level of support. The grey violin plot indicates the 95% highest posterior density (HPD) interval for 254 the tMRCA for the epidemic in the Americas (AM). Lineage F4 contains two identical ZIKV genomes 255 from the same patient. (b) A zoomed in version of the whole MCC tree showing the collection locations 256 of Miami-Dade sequences and whether they were sequenced from mosquitoes (numbers correspond to 257 trap locations in Fig. 1c). 95% HPD intervals are shown for the tMRCAs (c) The probability of ZIKV persistence after introduction for different R_0 . Persistence is measured as the number of days from initial 258 259 introduction of viral lineages until their extinction. Vertical dashed lines show the inferred mean 260 persistence time for lineages F1, F2 and B based on their tMRCA. (d) Total number of introductions 261 (mean with 95% CI) that contributed to the outbreak of 241 local cases in Miami-Dade County for 262 different R_0 .

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264 Figure 3 | Frequent opportunities for Zika virus introductions into Miami from the Caribbean. (a) 265 Reported ZIKV cases per country/territory from January to June, 2016 normalized by total population. (b) 266 The number of estimated travelers entering Miami during January to June, 2016 by method of travel. (c) 267 The number of travelers and the reported ZIKV incidence rate for the country/territory of origin were 268 used to estimate the proportion of infected travelers coming from each region with ZIKV in the Americas. 269 (d) The observed number of weekly travel-associated ZIKV cases in Florida were plotted with the 270 expected number of ZIKV-infected travelers (as estimated in panel c) coming from all of the Americas 271 (grey line) and the regional contributions (colored areas). (e) The countries visited by the 1,016 travel-272 associated ZIKV cases diagnosed in Florida.

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Figure 4 | Southern Florida has a high potential for *Aedes aegypti*-borne virus outbreaks. The estimated number of travelers per month (circles) entering Florida cities via flights and cruise ships were plotted with estimated relative *Ae. aegypti* abundance. Only cities receiving >10,000 passengers per month are shown. Relative *Ae. aegypti* abundance for every month is shown in Extended Data Fig. 1d.

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355 Methods

356 Ethical statement

357 This work was evaluated and approved by relevant Institutional Review Boards (IRB)/Ethics Review 358 Committees at The Scripps Research Institute (TSRI) and the US Army Medical Research Institute of 359 Infectious Diseases (USAMRIID) Office of Human Use and Ethics. This work was conducted as part of 360 the public health response in Florida and samples were collected under a waiver of consent granted by the Florida DOH Human Research Protection Program. The work received a non-human subjects research 361 designation (category 4 exemption) by the Florida DOH since this research was performed with leftover 362 363 clinical diagnostic samples involving no more than minimal risk. All samples were deidentified prior to 364 receipt by the study investigators.

365 Florida Zika virus case data

366 Weekly reports of international travel-associated and locally-acquired ZIKV infections diagnosed in 367 Florida were obtained from the Florida DOH mosquito-borne disease surveillance system⁴. Dates of symptom onset from the Miami transmission zones (Wynwood, Miami Beach, and Little River) 368 determined by the Florida DOH investigation process were obtained from the ZIKV resource website³⁵ 369 and daily updates³⁶. International travel-associated ZIKV case counts in the United States (outside of 370 Florida) were obtained from the CDC³⁷. The local and travel-associated ZIKV case numbers for Florida 371 were obtained from the Florida DOH. The one local ZIKV infection diagnosed in Duval County was 372 believed to have originated elsewhere in Florida. Therefore, this case is listed as "unknown origin" in Fig. 373 374 1b. In Fig. 3e, only the countries visited by 5 or more times by ZIKV-infected travelers diagnosed in Florida are shown. Countries with 5 or fewer visits were aggregated into an "other" category by region 375 376 (*i.e.*, Caribbean, South America, or Central America).

377 Clinical sample collection and RNA extraction

378 Clinical samples from locally-acquired ZIKV infections were collected from June 22 to October 11, 2016. 379 The Florida DOH identified persons with compatible illness and clinical samples were shipped to the 380 Bureau of Public Health Laboratories for confirmation by qRT-PCR and antibody tests following interim guidelines^{3,38-40}. Clinical specimens (whole blood, serum, saliva, or urine) submitted for analysis were 381 382 refrigerated or frozen at \leq -70°C until RNA was extracted. RNA was extracted using the RNAeasy kit 383 (QIAGEN), MagMAX for Microarrays Total RNA Isolation Kit (Ambion), or MagNA Pure LC 2.0 or 96 Systems (Roche Diagnostics). Purified RNA was eluted into 50-100 µL using the supplied elution 384 buffers, immediately frozen at \leq -70°C, and transported on dry ice. The Florida DOH also provided 385 386 investigation data for these samples, including symptom onset dates and, when available, assignments to 387 the zone where infection likely occurred (Supplementary Table 1).

388 Mosquito collection, RNA extraction, and entomological data analysis

389 24,351 Ae. aegypti and Ae. albopictus mosquitoes (sorted into 2,596 pools) were collected throughout 390 Miami-Dade County during June to November, 2016 using BG-Sentinel mosquito traps (Biogents AG). 391 Up to 50 mosquitoes of the same species and sex were pooled per trap. The pooled mosquitoes were 392 stored in RNAlater (Invitrogen), RNA was extracted using either the RNAeasy kit (QIAGEN) or 393 MagMAX for Microarrays Total RNA Isolation Kit (Ambion), and ZIKV RNA was detected by qRT-PCR targeting the envelope protein coding region⁴⁰ or the Trioplex qRT-PCR kit⁴¹. ZIKV infection rates 394 were calculated per 1,000 female Ae. aegypti mosquitoes using the bias-corrected maximum likelihood 395 estimate (MLE)⁴². Days of insecticide usage by the Miami-Dade Mosquito Control were inferred from the 396 zone-specific ZIKV activities timelines published by the Florida DOH³ 397

Relative monthly Ae. aegypti abundance 398

For the purpose of this study we used Ae. aegypti suitability maps from Kraemer et al.¹⁴ and derived 399 monthly estimates based on the statistical relationships between mosquito presence and environmental 400 correlates⁴³. Following Hwang *et al.*⁴⁴ we used a simple mathematical formula to transform the 401 probability of detection maps into mosquito abundance maps. In order to do so, we assumed P (Y=1) 402 403 where Y is a binary variable (presence/absence). Using a Poisson distribution X() to govern the 404 abundance of mosquitoes, the probability of not observing any mosquitoes can be related to the 405 probability of absence as: P(X=0)=P(Y=0). We used the following transformation to generate abundance 406 (λ) estimates per county in Florida:

$$e^{-\lambda} = P(Y = 0)$$
$$\lambda = -\log(P(Y = 0))$$
$$\lambda = -\log(1 - P(Y = 1))$$

407 We did not consider Ae. albopictus abundance in this study because 99.8% of mosquitoes collected in 408 Miami-Dade County were Ae. aegypti. Relative Ae. aegypti abundance in major U.S. cities presented in 409 Extended Data Fig. 8 was estimated as previously described²².

410 Zika virus quantification

411 ZIKV genome equivalents (GE) were quantified by qRT-PCR. At TSRI, ZIKV qRT-PCR was performed 412 as follows: ZIKV RNA standards were transcribed from the ZIKV NS5 region (8651-9498 nt) using the T7 forward primer (5' - TAA TAC GAC TCA CTA TAG GGA GA TCA GGC TCC TGT CAA AAC 413 414 CC - 3'), reverse primer (5' - AGT GAC AAC TTG TCC GCT CC - 3'), and the T7 Megascript kit 415 (Ambion). For qRT-PCR, primers and a probe targeting the NS5 region (9014-9123 nt) were designed 416 using the ZIKV isolate PRVABC59 (GenBank: KU501215): forward primer (5'- AGT GCC AGA GCT 417 GTG TGT AC - 3'), reverse primer (5' - TCT AGC CCC TAG CCA CAT GT - 3'), and FAM-fluorescent 418 probe (5' - GGC AGC CGC GCC ATC TGG T - 3'). The qRT-PCR assays were performed in 25 µl 419 reactions using the iScript One-step RT-PCR Kit for probes (Bio-Rad Laboratories Inc.) and 2 µl of 420 sample RNA. Amplification was performed at 50°C for 20 min, 95°C for 3 min, and 40 cycles of 95°C 421 for 10 s and 57°C for 10 s. Fluorescence was read at the end of the 57°C annealing-extension step. 10-fold 422 dilutions of the ZIKV RNA transcripts (2 µl/reaction) were used to create a standard curve for 423 quantification of ZIKV GE/µl of RNA. The lower limits of quantification are 4 GE/µl RNA, or at a cycle 424 threshold of \sim 36.

ZIKV GE were quantified at USAMRIID using the University of Bonn ZIKV envelope protein (Bonn E) 425 426 qRT-PCR assay⁴⁵. RNA standards were transcribed using an amplicon generated from a ZIKV plasmid 427 containing T7 promoter at the start of the 5' untranslated region (UTR). The plasmid was designed using 428 the ZIKV isolate BeH819015 (GenBank: KU365778.1) and the amplicon included nts 1-4348, which 429 covers the 5' UTR, C, prM, M, E, NS1, and NS2 regions. The qRT-PCR assays were performed in 25 ul reactions using the SuperScript III platinum One-step qRT-PCR Kit (ThermoFisher) and 2 µl of sample 430 RNA was used. Amplification was performed following conditions as previously described⁴⁵. 10-fold 431 432 dilutions of the ZIKV RNA transcripts (5 µl/reaction) were used to create a standard curve for 433 quantification of ZIKV GE/µl of RNA.

434 **Amplicon-based Zika virus sequencing**

ZIKV sequencing at TSRI was performed using an amplicon-based approach using the ZikaAsian V1 435

scheme, as described²⁴. This approach is similar to "RNA jackhammering" to sequence low-quality viral 436

samples developed by Worobey et al.⁴⁶. Briefly, cDNA was reverse transcribed from 5 µl of RNA using 437

SuperScript IV (Invitrogen). ZIKV cDNA (2.5 μ l/reaction) was amplified in 35 × 400 bp fragments from 438 two multiplexed PCR reactions using Q5 DNA High-fidelity Polymerase (New England Biolabs). The 439

amplified ZIKV cDNA fragments (50 ng) were prepared for sequencing using the Kapa Hyper prep kit
(Kapa Biosystems) and SureSelect XT2 indexes (Agilent). Agencourt AMPure XP beads (Beckman
Coulter) were used for all purification steps. Paired-end 251 nt reads were generated on the MiSeq using
the V2 500 cycle or V3 600 cycle kits (Illumina).

444 Trimmomatic was used to remove primer sequences (first 22 nt from the 5' end of the reads, which is the 445 maximum length of the primers used for the multiplexed PCR) and bases at both ends with Phred quality score $< 20^{47}$. The reads were then aligned to the complete genome of a ZIKV isolate from the Dominican 446 447 Republic, 2016 (GenBank: KU853012) using Novoalign v3.04.04 (www.novocraft.com). Samtools was used to sort the aligned BAM files and to generate alignment statistics⁴⁸. Snakemake was used as the 448 workflow management system⁴⁹. The code and reference indexes for the pipeline can be found at 449 https://github.com/andersen-lab/zika-pipeline. ZIKV-aligned reads were visually inspected using 450 Geneious v9.1.5⁵⁰ before generating consensus sequences. A minimum of $3 \times$ read-depth coverage, in 451 452 support of the consensus, was required to make a base call.

453 Enrichment-based Zika virus sequencing

454 ZIKV sequencing at USAMRIID was performed using a targeted enrichment approach. Sequencing 455 libraries were prepared using the TruSeq RNA Access Library Prep kit (Illumina) with custom ZIKV 456 probes. The set included 866 unique probes each of which was 80 nt in length (Supplementary Table 2a). The probes were designed to cover the entire ZIKV genome and to encompass the genetic diversity 457 458 present on GenBank on January 14, 2016. In total, 26 ZIKV sequences were used during probe design 459 (Supplementary Table 2b). Extracted RNA was fragmented at 94 °C for 0-60 s and each sample was 460 enriched separately using a quarter of the reagents specified in the manufacturer's protocol. Samples were 461 barcoded, pooled and sequenced using the MiSeq Reagent kit v3 (Illumina) on an Illumina MiSeq with a minimum of 2×151 bp reads. Dual indexing, with no overlapping indices, was used. 462

The random hexamer associated with read one and the Illumina adaptors were removed from the 463 sequencing reads using Cutadapt v1.9.dev1⁵¹, and low-quality reads/bases were filtered using Prinseq-lite 464 465 v0.20.3⁵². Reads were aligned to a reference genome (GenBank: KX197192.1) using Bowtie2 v2.0.6⁵³, duplicates were removed with Picard (http://broadinstitute.github.io/picard), and a new consensus was 466 v0.1.18⁴⁸ 467 generated using combination of Samtools and custom scripts а 468 (https://github.com/jtladner/Scripts/blob/master/reference-based assembly/consensus fasta.pv). Only 469 bases with Phred quality score ≥ 20 were utilized in consensus calling, and a minimum of $3 \times$ read-depth coverage, in support of the consensus, was required to make a call; positions lacking this depth of 470 471 coverage were treated as missing (*i.e.* called as "N").

472 Validation and comparison of sequencing methods

473 The consensus ZIKV sequences from FL01M and FL03M generated by sequencing 35×400 bp 474 amplicons on the MiSeq were validated using the following approaches: 1) sequencing the 35×400 bp amplicons on the Ion S5 platform (ThermoFisher), 2) sequencing amplicons generated using an Ion 475 AmpliSeq® (ThermoFisher) panel customly targeted towards ZIKV on the Ion S5 platform, and 3) 476 477 sequencing $5 \times 2,150$ -2,400 bp ZIKV amplicons on the MiSeq. For Ion library preparation, cDNA was 478 synthesized using the SuperScript VILO kit (ThermoFisher). ThermoFisher designed 875 custom ZIKV 479 primers to produce 75 amplicons of ~200 bp in two PCR reactions for use with their Ion AmpliSeq 480 Library Kit 2.0. The reagent FuPa was used to digest the modified primer sequences after amplification. 481 The DNA templates were loaded onto Ion 520 chips using the Ion Chef and sequenced on the Ion S5 with the 200 bp output (ThermoFisher). The 35×400 bp amplicons generated for the MiSeq as described 482 483 above were introduced into the Ion workflow using the Ion AmpliSeq Library Kit 2.0, but without 484 fragmentation. Primers to amplify 2,150-2,400 bp ZIKV fragments (Supplementary Table 2c) were kindly 485 provided by Shelby O'Connor, Dawn Dudly, Dave O'Connor, and Dane Gellerup (AIDS Vaccine 486 Research Laboratory, University of Wisconsin, Madison). Each fragment was amplified individually by

PCR using the cDNA generated above, Q5 DNA High-fidelity Polymerase, and the following 487 thermocycle conditions: 55 °C for 30 m, 94 °C for 2 m, 35 cycles of 94 °C for 15 s, 56 °C for 30 s, and 68 488 °C for 3.5 m, 68 °C for 10 m, and held at 4 °C until use. Each PCR product was purified using Agencourt 489 AMPure XP beads, sheared to 300 to 400 nt fragments using the Covaris S2 sonicator, indexed and 490 prepared for sequencing as described above, and sequenced using the MiSeq V2 500 cycle kit (paired-end 491 492 251 nt reads). Compared to the consensus sequences generated using 35×400 bp amplicons on the 493 MiSeq, there were no consensus-level mismatches in the coding sequence using any of the other three 494 approaches (Extended Data Table 2). There were, however, some mismatches in the 5' and 3' UTRs 495 (where the genomic RNA is heavily structured), likely a result of PCR bias and decreased coverage depth.

496 At least 95% of the ZIKV genome was covered from samples with as low as 4 and 9 GE/ μ l RNA from the 497 amplicon and enrichment approaches, respectively. These results are similar to our previously determined 498 clinical range of 10-16 ZIKV GE/ μ l RNA to achieve at least 95% genome coverage using our amplicon-499 based approach²⁴. On average, the amplicon-based sequencing approach covered 97% of the ZIKV 500 genome (\geq 3× read-depth) and the targeted enrichment approach covered 82% of the ZIKV genome from 501 clinical samples (Supplementary Table 2d).

502 **Phylogenetic analyses**

All published and available complete ZIKV genomes of the Asian genotype from the Pacific and the Americas were retrieved from GenBank public database as of December 2016. Public sequences (n=65) were codon-aligned together with ZIKV genomes generated in this study (n=39) using MAFFT⁵⁴ and inspected manually. The multiple alignment contained 104 ZIKV sequences collected between 2013 and 2016, from the Pacific (American Samoa, French Polynesia, and Tonga), Brazil, other South and Central Americas (Guatemala, Mexico, Suriname, and Venezuela), the Caribbean (Dominican Republic, Guadeloupe, Haiti, Martinique, and Puerto Rico), and the United States (Supplementary File 1).

510 In order to determine the temporal signal of the sequence dataset, a maximum likelihood (ML) phylogeny 511 was first reconstructed with PhyML⁵⁵ using the general time-reversible (GTR) nucleotide substitution 512 model and gamma distributed rates amongst sites⁵⁶ (Supplementary File 1), which was identified as the 513 best fitting model for ML inference by jModelTest2⁵⁷. Then, a correlation between root-to-tip genetic 514 divergence and date of sampling was conducted in TempEst⁵⁸.

Bayesian phylogenetic analyses were performed using BEAST v.1.8.4²⁵ to infer time-structured 515 phylogenies. We used an SDR06 nucleotide substitution model⁵⁹ with a non-informative continuous time 516 Markov chain reference prior (CTMC)⁶⁰ on the molecular clock rate. Replicate analyses using multiple 517 combinations of molecular clock and coalescent models were explored to select the best fitting model by 518 marginal likelihood comparison using path-sampling and stepping-stone estimation approaches⁶¹⁻⁶³ 519 (Extended Data Table 1b). The best fit model was a relaxed molecular clock along with a Bayesian 520 Skyline model⁶⁴. All the Bayesian analyses were run for 30 million Markov chain Monte Carlo steps, 521 sampling parameters and trees every 3000 generations (BEAST XML file and MCC tree available in 522 523 Supplementary File 1). Support values for all nodes are embedded in the phylogenetic tree files (Supplementary File 1). Tree visualizations were generated with baltic (github.com/blab/baltic). 524

525 The travel-associated ZIKV genomes add to the Caribbean dataset, but do not directly influence our conclusions about the source of ZIKV introductions into Florida.

527 Expected number and distribution of local cases from Zika virus importations

528 We used branching process theory^{65,66} to generate the offspring distribution (subsequent local cases) that

- 529 is expected from a single introduction. The offspring distribution L is modelled with a negative binomial
- 530 distribution with mean R_0 and over-dispersion parameter k. The total number of cases j that is caused by a
- single importation (including the index case) after an infinite time 67 has the following form:

$$L = \frac{\Gamma(kj+j-1)}{\Gamma(kj)\,\Gamma(j+1)} \frac{(\frac{R_0}{k})^{j-1}}{(1+\frac{R_0}{k})^{kj+j-1}}$$

The parameter k represents the variation in the number of secondary cases generated by each case of ZIKV⁶⁵. In the case of vector borne diseases, local heterogeneity is high due to a variety of factors such as mosquito population abundance, human to mosquito interaction, and control interventions^{68–73}. Here, we assumed high heterogeneity (k=0.1) following previous estimates for vector borne diseases⁶⁶. This distribution *L* is plotted in Extended Data Fig. 4a. For the following, we took a forward simulation approach, drawing random samples from this distribution. All estimates were based on 100,000 random simulations.

We used this formula to estimate the probability of observing 241 local cases in Miami-Dade County alongside 320 travel-associated cases. We approached this by sampling 320 introduction events from *L* and calculating the total number of local cases in the resulting outbreak (Extended Data Fig. 4b). We also calculated the likelihood of observing 241 local cases in the total outbreak (Extended Data Fig. 4c), finding that the MLE of R_0 lies between 0.35 and 0.55. As a sensitivity analysis, we additionally modelled introductions with the assumption that only 50% of travelers were infectious at time of arrival into Miami-Dade County, resulting in an MLE of R_0 of 0.45–0.8.

546 We further used this formula to address the probability of observing 3 distinct genetic clusters (F1, F2 and 547 F3) representing 3 introduction events in a sample of 27 ZIKV genomes from Miami-Dade County. We 548 approached this by sampling introduction events until we accumulated 241 local cases according to L, arriving at N introduction events with case counts $(j_1, j_2, ..., j_N)$. We then sampled 27 cases without replacement from $(j_1, j_2, ..., j_N)$ following a hypergeometric distribution and recorded the number of 549 550 distinct clusters drawn in the sample. We found that higher values of R_0 resulted in fewer distinct clusters 551 within the sample of 27 genomes (Extended Data Fig. 4d). We additionally calculated the likelihood of 552 553 sampling 3 distinct genetic clusters in 27 genomes (Extended Data Fig. 4e), finding an MLE estimate of 554 R_0 of 0.7–0.9. Additionally, as a sensitivity analysis we modelled a preferential sampling process in which larger clusters are more likely to be drawn from than smaller clusters. Here, we used a parameter α that enriches the hypergeometric distribution following $(j_1^{\alpha}, j_2^{\alpha}, \dots, j_N^{\alpha})$. In this case, we found an MLE 555 556 557 estimate of R_0 of 0.5–0.9.

558 Using the overlap of estimates of R_0 from local case counts (0.35–0.8) and genetic clusters (0.5–0.9), we arrived at a 95% uncertainty range of R_0 of 0.5–0.8. As an additional sensitivity analysis, we incorporated 559 560 under-reporting in which either 50% of travel-associated cases and 25% of local cases are reported or in 561 which 10% of travel-associated cases and 5% of local cases are reported. We find differential reporting of 562 travel and local cases results in increased mean R₀ estimates when comparing counts of travel-associated to local cases (Extended Data Figure 4f-g). Additionally, we find that under-reporting increases estimates 563 of R_0 from the sampling analysis (Extended Data Figure 4h-i). Thus, moderate under-reporting is 564 565 consistent with R_0 estimates of ~0.8.

566 We additionally perform birth-death stochastic simulations assuming a serial interval with mean 20 567 days¹⁵. We record the number of stochastic simulations still persisting after a particular number of days 568 for different values of R_0 (Fig. 2c).

569 Zika virus incidence rates

570 Weekly suspected and confirmed ZIKV case counts from countries and territories within the Americas

- 571 with local transmission (January 1 to September 18, 2016) were obtained from the Pan American Health
- 572 Organization (PAHO)³⁰. In most cases, the weekly case numbers per country were only reported in bar
- 573 graphs. We contacted PAHO multiple times with the hope of gaining access to the raw data included in

the bar graphs, but our requests were unfortunately denied. Therefore we used WebPlotDigitizer v3.10 (http://arohatgi.info/WebPlotDigitizer) to estimate the numbers. We compared the actual ZIKV case numbers reported in Ecuador⁷⁴ (only country with available raw data and reported cases > 10 per week) to our estimates from the PAHO bar graphs and found that the WebPlotDigitizer was ~99% accurate (Extended Data Fig. 5a-b).

579 Country and territory total population sizes to calculate weekly and monthly ZIKV incidence rates were 580 also obtained from PAHO⁷⁵. Incidence rates calculated from countries and territories in the Americas 581 during January to June, 2016 (based on the earliest introduction time estimates until the first known cases) 582 were used as an estimate for infection likelihood to investigate sources of ZIKV introductions.

583 Airline and cruise ship traffic

584 To investigate whether the transmission of ZIKV in Florida coincides with travel patterns from ZIKV 585 endemic regions, we obtained the number of passengers arriving at airports in Florida via commercial air 586 travel. We collated flight data from countries and territories in the Americas with local ZIKV 587 transmission between January and June, 2016 (based on the earliest introduction time estimates until the 588 first known cases, Supplementary Table 1b), arriving at all commercial airports in Florida. The data were 589 obtained from the International Air Transportation Association, which collects data on an estimated 90% of all passenger trips worldwide. Nelson et al.²⁸ previously reported flight data from 33 countries with 590 ZIKV transmission entering major United States airports during October 2014 through September 2015, 591 592 which we used to assess the potential for ZIKV introductions outside of Florida.

593 Schedules for cruise ships visiting Miami, Port Canaveral, Port Everglades, Fort Lauderdale, Key West, 594 Jacksonville (all in Florida), Houston, Galveston (both in Texas), Charleston (South Carolina) and New 595 Orleans (Louisiana) ports in the year 2016 were collated from www.cruisett.com and confirmed by crossreferencing ship logs reported by Port of Miami and reported ship schedules 596 from 597 Scheduled www.miamidade.gov/portmiami/. cruise ship capacities were extracted from 598 www.cruisemapper.com. Every country/territory with ZIKV transmission visited by a cruise ship 10 days 599 (the approximate mean time to ZIKV clearance in human blood [*i.e.*, the infectious period])⁷⁶ prior to arrival was counted as contributing the ship's capacity worth of passengers to Miami to the month of 600 arrival (Supplementary Table 1b). While the air traffic was based on the reported number of travelers, we 601 estimated the sea traffic by ship capacity. Lee and Ramdeen⁷⁷ reported that the average occupancy of 602 cruise ships traveling to the Caribbean Islands exceeded 100% in 2011, and according to the Florida-603 Caribbean Cruise Association⁷⁸, it remained >100% in 2015. Occupancy data for 2016 was not available 604 at the time of publication, but we assumed that it was also near 100%. 605

606 Expected number of travelers infected with Zika virus

607 We estimated the expected number of travelers entering Miami who were infected with ZIKV (λ) by

608 using the total travel capacity (C) and the likelihood of ZIKV infection (infections (I) per person (N)) 609 from each country/territory (i):

$$\lambda = \sum_{i} C_{i} \frac{I_{i}}{N_{i}}$$

610 We summed the number of expected infected travelers from each country/territory with ZIKV 611 transmission by region and travel method (flights or cruises). The number of ZIKV cases reported by each 612 country are likely under-estimates in part because the majority of ZIKV infections are asymptomatic^{2,31}. 613 We normalized some of the potential reporting variances between countries by reporting the data as the

relative proportion of infected travelers (Fig. 3c, Extended Data Fig. 7a) and as the absolute number of

615 infected travelers (Fig. 3d, Extended Data Fig. 7b, Supplementary Table 1b) from each region. We also

616 accounted for potential reporting biases with incidence rates by using ZIKV attack rates (*i.e.*, proportion

617 infected before epidemic burnout) to estimate peak transmission intensity. Attack rates were calculated

618 using a susceptible-infected-recovered (SIR) transmission model derived from seroprevalence studies

and environmental factors as described⁷⁹. Using attack rates as an estimate of infection likelihood, we predict that ~60% of the infected travelers entering Miami came from the Caribbean (Extended Data 7b),

620 predict that ~60% of the infected travelers entering Miami came from the Caribbean (Extended Data 7b), which is in agreement with our methods using incidence rates of ~60-70% (Fig. 3c). A list of countries

and territories used in these analyses can be found in Supplementary Table 1b.

623 Maps

624 The maps presented in our figures were generated using Matplotlib⁸⁰ and ESRI basemaps 625 (www.esri.com/data/basemaps). The software and basemaps are open source and "freely available to 626 anyone".

627 Data availability

All ZIKV sequencing data is available under the NCBI BioProjects PRJNA342539 and PRJNA356429.
 Individual sample GenBank access numbers are listed in Supplementary Table 1a. All other data is available in the Extended Data, Supplemental Information, or upon request.

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634 Extended Data

635 Extended Data Fig. 1 | Miami-Dade mosquito surveillance and relative Aedes aegypti abundance. (a) Mosquito surveillance data reported from June 21 to November 28, 2016 was used to evaluate the risk of 636 ZIKV infection from mosquito-borne transmission in Miami. A total of 24,306 Ae. aegypti and 45 Ae. 637 638 *albopictus* were collected. Trap nights are the total number of times each trap site was used and the trap 639 locations are shown in Fig. 1d (some "Other Miami" trap sites are located outside of mapped region). Up 640 to 50 mosquitoes of the same species and trap night were pooled together for ZIKV RNA testing. The 641 infection rates were calculated using a maximum likelihood estimate (MLE). None of the Ae. albopictus 642 pools contained ZIKV RNA. (b) The number of weekly ZIKV cases (based on symptoms onset) was 643 correlated with mean Ae. aegypti abundance per trap night determined from the same week and zone 644 (Spearman r = 0.61). This suggests that when the virus is present, mosquito abundance numbers alone 645 could be used to target control efforts. (c) Insecticide usage, including truck and aerial adulticides and 646 larvacides, by the Miami-Dade Mosquito Control in Wynwood (left) and Miami Beach (right) was 647 overlaid with Ae. aegypti abundance per trap night to demonstrate that intense usage of insecticides may 648 have helped to reduce local mosquito populations. (d) Relative Ae. aegypti abundance for each Florida 649 county and month was estimated using a multivariate regression model, demonstrating spatial and 650 temporal heterogeneity for the risk of ZIKV infection.

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652 Extended Data Fig. 2 | Maximum likelihood tree and root-to-tip regression of Zika virus genomes 653 from Pacific islands and the epidemic in Americas. (a) Maximum likelihood tree of publicly available 654 ZIKV sequences and sequences generated in this study (n=104), tips are coloured by location, labels in 655 bold indicate sequences generated in this study, Florida clusters F1-F4 are indicated by vertical lines to 656 the right of the tree. Bootstrap support values are shown at key nodes. All other support values can be 657 found in Supplementary File 1. (b) Linear regression of sample tip dates against divergence from root based on sequences with known collection dates estimates an evolutionary rate for the ZIKV phylogeny 658 of 1.10×10^{-3} nucleotide substitutions/site/year (subs/site/yr). This is consistent with BEAST analyses 659 using a relaxed molecular clock and a Bayesian Skyline tree prior, the best-performing combination of 660 clock and demographic model according to marginal likelihood estimates (Extended Data Table 1c), 661 which estimated an evolutionary rate of 1.21×10^{-3} (95% highest posterior density: $1.01 - 1.43 \times 10^{-3}$) 662 663 subs/site/vr (Extended Data Table 1a). These values are in agreement with previous estimates calculated 664 based on ZIKV genomes from Brazil⁶.

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666 Extended Data Fig. 3 | Molecular clock dating of Zika virus clades. Maximum clade credibility 667 (MCC) tree of ZIKV genomes collected from Pacific islands and the epidemic in Americas (n=104). 668 Circles at the tips are colored based on origin location. Clade posterior probabilities are indicated by 669 white circles filled with black relative to the support. A posterior probability of 1 fills the entire circle 670 black. The grey violin plot indicates the 95% highest posterior density (HPD) interval for the tMRCA of 671 the American epidemic. We estimated that the tMRCA for the ongoing epidemic in the Americas 672 occurred during October, 2013 (node AM, Extended Table 1, 95% HPD: August, 2013-January, 2014), 673 which is consistent with previous analysis based on ZIKV genomes from Brazil⁶.

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Extended Data Fig. 4 | Estimation of basic reproductive number and number of introductions in 675 676 Miami-Dade County. (a) Probability distribution of estimated total number of cases caused by a single introduction (excluding the index case) for different values of R_0 . (b) Mean and 95% CI for total number 677 678 of local cases caused by 320 introduction events (*i.e.*, travel-associated cases diagnosed in Miami-Dade 679 County) for different values of R_0 and for different assumptions of proportion of infectious travelers. (c) 680 Log likelihood of observing 241 local cases in Miami-Dade County with 320 introduction events for different values of R_0 along with 95% maximum likelihood estimate (MLE) bounds on R_0 . (d) Mean and 681 95% uncertainty interval for total number of distinct phylogenetic clusters observed in 27 sequenced 682 683 ZIKV genomes from human cases diagnosed in Miami-Dade County for different values of R_0 and for

684 different assumptions of sampling bias, from $\alpha=1$ (no sampling bias) to $\alpha=2$ (skewed toward 685 preferentially sampling larger clusters). (e) Log likelihood of observing 3 clusters (*i.e.*, ZIKV lineages F1, 686 F2, and F4, Fig. 2a) in 27 sequenced cases for different values of R_0 along with 95% MLE bounds on R_0 . 687 (f) Mean and 95% CI for total number of local cases caused by 320 observed travel-associated cases with 688 travel-associated vs local reporting rates of 50%/25% and 10%/5%. This assumes 50% of travelers are infectious. (g) Log likelihood of observing 241 local cases with 320 introduction events for different 689 values of R_0 along with 95% MLE bounds on R_0 with travel-associated vs local reporting rates of 690 50%/25% and 10%/5%. (h) Mean and 95% uncertainty interval for total number of distinct phylogenetic 691 692 clusters observed in 27 sequenced ZIKV genomes for different values of R_0 and for assumptions of local 693 reporting rate of 5% and 25%. This assumes preferential sampling (α =2). (i) Log likelihood of observing 3 clusters in 27 sequenced cases for different values of R_0 along with 95% MLE bounds on R_0 with local 694 reporting rate of 5% and 25%. At 5% local reporting rate, 0 of the 100,000 replicates for all R_0 values 695 696 showed 3 clusters.

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698 Extended Data Fig. 5 | Weekly reported Zika virus case numbers and incidence rates in the Americas. (a) Most ZIKV case numbers reported by PAHO³⁰ were only available as bar graphs (raw data 699 700 was not made available to us at the time of request). Therefore we used the WebPlotDigitizer to estimate 701 the weekly case numbers from the PAHO bar graphs. ZIKV cases reported from Ecuador was the only data set to include a link to the actual case numbers that also had >10 cases per week⁷⁴. To validate the 702 703 WebPlotDigitizer, we compared the weekly reported case numbers from Ecuador to our estimates. (b) 704 The reported and estimated case numbers were strongly correlated (Spearman r = 0.9981). The 705 WebPlotDigitizer was used to estimate the ZIKV case numbers for all subsequent analysis. (c) ZIKV 706 cases (suspected and confirmed) and (d) incidence rates (normalized per 100,000 population) are shown 707 for each country or territory with available data per epidemiological week from January 1 to September 708 18, 2016. (e) Each country or territory with available data is colored by its reported ZIKV incidence rate 709 from January to June, 2016 (the time frame for analysis of ZIKV introductions into Florida).

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711 Extended Data Fig. 6 | Cruise and flight traffic entering Miami from regions with Zika virus 712 transmission. The estimated number of passengers entering Miami, by either (a) cruises or (b) flights, 713 from each country or territory in the Americas with ZIKV transmission per month (left panel). The center 714 map and inset show the cumulative numbers of travelers entering Miami during January to June, 2016 715 (the time frame for analysis of ZIKV introductions into Florida) from each country or territory per method 716 of travel. (c) The total traffic (*i.e.* cruises and flights) is shown entering Miami per month.

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718 Extended Data Fig. 7 | Expected number of Zika virus infected travelers from the Caribbean is 719 correlated with the total observed number of travel-associated infections. (a) In order to account for 720 potential biases in ZIKV reporting accuracies, we also estimated the proportion of infected travelers using 721 projected ZIKV attack rates⁷⁹ (*i.e.* predicted proportion of population infected before epidemic burnout). About 60% of the infected travelers are expected to have arrived from the Caribbean, similar to our 722 723 results using incidence rates (Fig. 3c). (b) The expected number of travel-associated ZIKV cases were 724 estimated by the number of travelers coming into Miami from each country/territory (travel capacity) and 725 the in-country/territory infection likelihood (incidence rate per person) per week. The expected travel 726 cases were summed from all of the Americas (left), Caribbean (left center), South America (right center), 727 and Central America (right) and plotted with the observed travel-associated ZIKV cases. Numbers in each 728 plot indicate Spearman correlation coefficients. Negative Spearman r coefficients indicated a negative 729 correlation between the number of expected and observed travel cases.

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Fig. 8 | Greater early season potential for Zika virus introductions into Miami. The monthly cruise ship and airline²⁸ capacity from countries/territories with ZIKV transmission for the major United States travel hubs (shown as circle diameter) with monthly potential *Ae. aegypti* abundance (circle color), as previously estimated²². The abundance ranges were chosen with respect to the May-Oct Miami

- 735 mean: "None to low" (<2%), "Low to moderate" (2-25%), "Moderate to high (25-75%), and "High"
- (>75%). Mosquito-borne transmission is unlikely in the "None to low" range. Cruise capacities from
 Houston and Galveston, Texas were combined.
- 738

Extended Data Table 1 | (a) Time of the most recent common ancestor and evolutionary rate and (b) Model selection to infer time-structured phylogenies.

- HPD, highest posterior density. Dates listed as proportion of days elapsed with a year. Clades refer to Fig.
 2a.
- 743

744 Extended Data Table 2 | Validation of sequencing results.

- ^a Compared to the consensus genomes generated by sequencing 35×400 bp amplicons on the MiSeq.
- ^b Amplicons produced using Ion AmpliSeq and 875 custom ZIKV primers.
- 747 NGS, next-generation sequencing; UTR, untranslated region; CDS, coding sequence.
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