Reliable estimates of Japanese encephalitis virus evolutionary rate supported

by a formal test of temporal signal

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Abstract

Japanese encephalitis virus (JEV) is a positive-strand RNA virus, that causing

encephalitis in humans, which is highly spread in East and Southeast Asia,

Western Pacific countries and Australia. The main vector of JEV is *Culex* spp.

(mainly *C. tritaeniorhynchus*). The intraspecies structure of JEV comprises

five genotypes (GI-GV). In the previous studies, the substitution rate and the

age of the JEV most recent common ancestor (MRCA) based on complete

open reading frame (ORF) / whole genome sequences were reported, however,

the estimates are significantly different between some of the studies. Most

importantly, the temporal signal in nucleotide data as the key aspect of the

molecular clock application was evaluated by neither formal nor informal

tests. In February 2023, we found 325 non-recombinant nucleotide sequences

of ORF with known collection dates. The purpose of our study is the

evaluation of temporal signal in the JEV ORF nucleotide data set using the

Bayesian approach with the subsequent re-estimation of substitution rate and

the age of the JEV common clade (GI-GV) and two JEV genotypes (GI and

GIII) characterized by the number of sequences enough for reliable estimates.

Results showed the presence of temporal signal in all analyzed data set. The

substitution rate of the JEV GI-GV was 2.41×10^{-4} (95% highest posterior

density (HPD) interval, 1.67×10^{-4}–3.14×10^{-4}) or approximately 2.5

nucleotide substitutions per ORF (10,296 nucleotides) per year that is one the

lowest substitution rate among the species closely related to JEV and other

well-studied members of mosquito-borne flaviviruses. The substitution rate of

GI and GIII evaluated as 4.13×10^{-4} (95% HPD interval, 3.45×10^{-4}–4.82×10^{-4}) and 6.17×10^{-5} (95% HPD interval 3.84×10^{-5}–8.62×10^{-5}), accordingly, with the ages of clades of 153 (95% HPD, 87–237) and 216 (95% HPD, 139–317), respectively. The root height of JEV is 1234 years (95% HPD, 407–2333). To the best of our knowledge, this study provides the first reliable substitution rate assessment of the global clade of JEV embracing all five genotypes.

Keywords: substitution rate, molecular clock, Orthoflavivirus, phylogeny, RNA virus, evolution

1. Introduction

Japanese encephalitis virus belongs to the species *Orthoflavivirus japonicum,* family *Flaviviridae* (Postler et al., 2023). The JEV genome consists of single-stranded positive-sense RNA with a length of about 10,965 nucleotides (nt). The genome encompasses a 10,296 nt single open reading frame (ORF), encoding a polyprotein, which includes three structural proteins: capsid (C), prosecutor membrane (prM) and envelope (E) proteins, and seven non-structural ones: NS1, NS2a, NS2b, NS3, N4a, NS4b, NS5 (Xu et al., 2023), (Lindenbach, 2007).

Mosquitoes are the vectors for JEV, in particular, *Culex tritaeniorhynchus* and *Culex annulirostris*, which were recognized as principal ones (Van den Eynde et al., 2022). The virus is zoonotic, and it is transmitted in most cases between pigs or birds by mosquitoes, however, Ricklin et al. (2016) described the direct way of virus transmission between two pigs via oronasal secretions. Humans, cattle and horses are dead-end hosts, due to insufficient viremia to infect mosquitoes. When these hosts are infected, encephalitis can develop with fever, convulsions, coma and even death (Van den Eynde et al., 2022), (Ladreyt et al., 2019).

Approximately 3 billion people live in JEV epidemic areas according to the World Health Organization data (https://www.who.int/news-room/fact-sheets/detail/japanese-encephalitis). JEV is the major causative agent of human virus encephalitis disease in Asia. Now the virus is endemic in Australia and Asia,

including 24 countries, namely China, Korea, Indonesia, Malaysia, Japan and even

Pakistan (Van den Eynde et al., 2022), (Mackenzie, 2005). Additionally, virus was

registered outside the endemic areas in Italy (Ravanini et al., 2012) and Angola

(Simon-Loriere et al., 2017).

Today, between 30,000 (Van den Eynde et al., 2022) and 69,000 (Auerswald

et al., 2021) cases registered annually according to different sources, but it is

considered as underestimated values due to insufficient monitoring and diagnosis.

The annual loss is reckoned as 709,000 disability-adjusted life years (Auerswald et

al., 2021). According to (McGuinness et al., 2023) JEV causes 100, 000 cases and

25, 000 deaths annually. The recent systematic review (Cheng et al., 2022) reported

that the case fatality rate of JEV in 2000–2018 was 14% (95% credible interval, 11–

17) and demonstrated the trend to decline in comparison with the previous periods.

Despite the fact that effective and safe vaccines have been developed (Vannice

et al., 2021), JEV is still one of the significant agents causing viral encephalitis in

the territory of Asia. Thus, to monitor and control this disease, it is important to

scrutinize evolutionary driving forces, which determine virus spread and

pathogenicity.

Currently, the JEV substitution rate estimates and the time to the most recent

common ancestor (MRCA) of the virus differ between the studies. Moreover, it is

important to underline that none of the observed works have performed the

evaluation of temporal signal in datasets prior to molecular clock application, and,

therefore, the reliability of the reported estimates is questionable.

The aim of our study is the reliable assessment of substitution rates and ages

of JEV and its genotypes using the Bayesian phylogenetic approach. For this

purpose, we used 325 complete ORF nucleotide sequences of all five JEV genotypes

with known collection dates available at the start of a study (January 2023).

Evolutionary rate and divergence times obtained were compared with the previous

estimates of JEV evolutionary rate and the rate of virus species, closely related to

JEV, and other well-studied mosquito-borne flaviviruses. Bayesian evaluation of

temporal signal makes reported results the first reliable assessment of JEV

substitution rate.

2. Materials and methods

2.1 Nucleotide dataset

Initially, 440 nucleotide sequences of complete or near complete genomes of

JEV were found in GenBank using BLAST (http://blast.ncbi.nlm.nih.gov/Blast.cgi).

Clones, vaccine and chimeric strains were removed from the dataset and then

collection dates were identified for 367 strains. Recombinant sequences detected by

RDP5 (Martin et al., 2021) were removed from the alignment. Thus, our research

was conducted using 325 nucleotide sequences of a polyprotein gene.

Next, sequence alignment was done with MAFFT v.7.490 (Katoh and

Standley, 2013). After that, the alignment was visualized in AliView v.1.24 (Larsson,

2014) to exclude ORFs with stop-codons. Non-coding regions were cut off using the

NCBI reference sequences (NC_001437.1).

JEV genotypes were determined in preliminary phylogenetic analysis in IQ-

TREE v.1.6.12 using the tree topology reported by (Takhampunya et al., 2011). It

was revealed that only GI and GIII have a sufficient number of sequences to obtain

reliable statistics in the further independent Bayesian phylogenetic analyses.

Therefore, we subsequently analyzed three data sets — the common clade of JEV

(JEV GI-GV), GI and GIII.

2.2 Temporal signal assessment

Temporal structure in each data sets (JEV GI-GV, GI and GIII) was assessed

with the Bayesian evaluation of temporal signal (BETS) (Duchene et al., 2020) in

BEAST v.2.6.7 (Bouckaert et al., 2019).

BETS involves the comparison of time-stamped (heterochronous) and

isochronous models (M_{het} and M_{iso}) by the difference between log marginal

likelihoods. The log marginal likelihoods for each model were calculated using the

path sampling estimator implemented in the MODEL_SELECTION package from

the BEAST2 Package Manager. The number of steps was set as 200 and the lengths

of MCMC for each step were determined so that the product of the number of steps
and MCMC length equaled the length of the MCMC during the main phylogenetic
analysis in BEAST with an effective sample size (ESS) at least of 200. Priors on
model parameters (molecular clock, substitution model, population model) were
chosen similar to those used in the main analysis with the exception of the tree prior
— in this case, we replaced the Bayesian skyline with the coalescent constant
population with a proper prior (i.e., integrated to 1) on effective population size
(Table 1):

Table 1. Priors used in the Bayesian evaluation of temporal signal (BETS) in all
three data sets (JEV GI-GV, GI, GIII)

Model	Parameter		Prior
Strict clock	Rate		Uniform(0.0, 1.0)
Optimized relaxed clock	Rates		LogNormal(M = 1.0, S = 0.2)
	Sigma		$\Gamma(\alpha = 0.5396, \beta = 0.3819)$
	Mean		$\Gamma(\alpha = 0.1, \beta = 100)$
Rate heterogeneity	Shape of Γ distribution		Exp(M = 1.0)
	Proportion of invariant sites		Uniform(0.0, 1.0)
Substitution model	GTR rates	Transition (A-G)	$\Gamma(\alpha = 0.05, \beta = 20)$
		Transversions	$\Gamma(\alpha = 0.05, \beta = 10)$
Tree prior (coalescent constant size)	Effective population size		Exp(M = 100)

Besides, BETS allows for a simultaneous and coherent assessment of the
entire model, including the molecular clock. Therefore, we included a strict clock
(SC) and an optimized relaxed clock (ORC) models into the comparison (Douglas
et al., 2021).

The difference between log marginal likelihoods of competitive models (i.e.,
log Bayes factor) of at least 5.0 indicates very strong support, a value of 3 means
strong support and a value of 1 is considered to be positive evidence (Kass and
Raftery, 1995).

Also, for comparison purposes, we conducted widely used root-to-tip (RTT)
regression, which is an informal test with well-known limitations: data points are

not independent and strict clocklike behavior is assumed. RTT statistics were

computed with Clockor2 (Featherstone et al., 2023). The trees for Clockor2 were

reconstructed in IQ-TREE v.1.6.12 with the best-fit model determined for each data

set using ModelFinder (Kalyaanamoorthy et al., 2017).

2.3 Dated phylogeny reconstruction and substitution rate evaluation

Bayesian phylogenetic analysis using JEV-common, GI and GIII data sets was

performed in BEAST v.2.6.7. Selecting substitution model was based on the BIC

previously assessed by ModelFinder in IQ-TREE. Molecular clock models were

compared based on the log MLs calculated on the step of BETS. As a tree prior, we

specified a flexible non-parametric Bayesian Skyline model. The lengths of MCMC

in each run were chosen so that the ESS associated with estimated parameters

reached at least 200 samples. The inspection of model parameters and MCMC

convergence was carried out in Tracer v.1.7.1 (Rambaut et al., 2018). The burn-in

thresholds of MCMC chains were defined individually depending on the chain

behavior. Most clade credible (MCC) trees were annotated in TreeAnnotator

(Bouckaert et al., 2019). Due to high computational cost, BEAST was run using a

BEAGLE library (Ayres et al., 2012) on a GPU (NVIDIA GeForce RTX 3070).

3. Results and discussion

3.1 Degree of clocklike behavior of the common JEV dataset (GI-GV) and

GI and GIII genotypes

In the first step, using three JEV data sets (GI-GV, GI, GIII) we conducted

RTT regression to assess to which degree data behave in a clock-like fashion

(Figure 1).

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Figure 1. Root-to-tips (RTT) regression analysis of three JEV data sets; X-axes represent
calendar years, Y-axes — distances from a root to tips.

JEV GI-GV and GIII data sets showed R^2 values about zero (0.04 and 0.12

with a negative slope, respectively) indicating the weak clocklike behavior of the

data which can be explained by the presence of a large amount of rate variation

between lineages of the trees (95% HPD of CV, 1.04–1.62 for GI-GV and 1.29–2.4

for GIII). GI, on the contrary, demonstrated a stronger temporal structure with R^2 of

0.79 and the lowest CV values (95% HPD, 0.43–0.68) (Table 2).

Table 2. Comparison of the determination coefficients inferred in root-to-tip
regression analysis and the coefficient of variation values.

Dataset	Number of samples	Sampling window, years	Determination coefficient (R^2) inferred in RTT analysis [1]	95% HPD of the coefficient of rate variation [2]
GI-GV	325	87	0.04	1.04–1.62
GI	134	45	0.79	0.43–0.68
GIII	171	87	0.12	1.29–2.4

173 [1] Root-to-tip (RTT) regression analysis was performed in Clockor2;
174 [2] 95% HPDs were inferred in BEAST analysis.
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Importantly, the low R^2 values obtained should not be considered as the

absence of a temporal signal. The point is that the RTT regression analysis appeared

to be uninformative when the data were sampled over a relatively narrow sampling

window (GI-GV data set) and there is large rate variation among lineages (GIII data

set) (Duchene et al., 2020). Therefore, we applied BETS — a formal test of temporal

signal — to assess the statistical support for including sampling times in a Bayesian

analysis.

3.2 The presence of temporal signal in all JEV data sets

BETS displayed very strong support (log Bayes factor > 5) for M_{het} over M_{iso}

in each pair of competitive models (Table 3) with the lowest log Bayes factor of 47.8

for the GIII data set (ORC_{het} relative to ORC_{iso}) which possessed the largest amount

of rate variation among branches (Table 2). Despite the last fact, in the log scale,

such BF values can be interpreted as overwhelming support for including sequence

sampling times into the model (Table 3).

Table 3. Results of the Bayesian test of temporal signal in Japanese
encephalitis virus data sets

Dataset	Molecular clock [1]	Log marginal likelihood	Log Bayes factor [2]
GI-GV	ORC_{iso}	−113559.93	140.38
	ORC_{het}	−113419.55	0
	SC_{iso}	−115098.41	1678.86
	SC_{het}	−114993.90	1574.34
GI	ORC_{iso}	−54633.51	129.02
	ORC_{het}	−54504.49	0
	SC_{iso}	−55182.96	678.48
	SC_{het}	−54734.24	229.75
GIII	ORC_{iso}	−46598.66	47.80
	ORC_{het}	−46550.86	0
	SC_{iso}	−47491.94	941.08
	SC_{het}	−47515.01	964.14

[1] ORC—optimized relaxed clock, SC—strict clock; iso—isochronous model, het—
heterochronous model;
[2] Best-performing model has a log Bayes factor value of 0.

The comparison of the SC_{het} and ORC_{het} models also showed very strong

evidence in favor of ORC in each dataset. The lowest log Bayes factor of 229.75

was detected for, as expected, GI which had the highest R^2 in the RTT regression

analysis.

Thus, all three data sets passed the Bayesian test of temporal signal and were

analyzed with ORC in the next step of phylogenetic analysis.

3.3 Substitution rates

The substitution rate of the common JEV GI-GV clade was 2.41×10^{-4} nucleotide substitutions per site per year (s/s/y) (95% HPD, 1.67×10^{-4}–3.14×10^{-4} s/s/y) or approximately 2.5 nucleotide substitutions per ORF (10,296 nucleotides) per year. GI has the highest rate of evolution: 4.13×10^{-4} s/s/y (95% HPD, 3.45×10^{-4}–4.82×10^{-4} s/s/y) or approximately 4.2 nucleotide substitutions per ORF per year. The lowest substitution rate was obtained for GIII: 6.17×10^{-5} s/s/y (95% HPD, 3.84×10^{-4}–8.63×10^{-4} s/s/y) or approximately 0.6 nucleotide substitutions per ORF per year. Notably, the 95% HPD intervals of estimates do not overlap (Figure 2).

Figure 2. Substitution rates of common JEV clade (GI-GV), genotypes I (GI) and III (GIII). Upper and lower ends of the boxes are 95 % highest posterior density interval borders.

3.3.1 Comparing the substitution rates of JEV yielded in our analysis with estimates reported previously

We compared the JEV substitution rate values obtained in our study with the previous estimates based on the analysis of complete ORF gene sequences (Figure 3).

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Figure 3. Comparison of the Japanese encephalitis virus (JEV) substitution rate evaluated
in our study based on complete open reading frame nucleotide sequences and the JEV substitution
rates according to the literature. The vertical bars represent 95% highest posterior density intervals,
circles—median or mean values. The X-axis labels are references to appropriate studies, the
methods on temporal signal evaluation applied are indicated above the X-axis: BETS—Bayesian
evaluation of temporal signal, RTT—root-to-tips regression, *—temporal signal was not assessed.

First of all, we found only one study where at least the degree of clock-like

behavior was assessed with informal RTT (Li et al., 2023) when analyzing GI.

Therefore, the substitution rate values reported in the literature should be treated

with caution, especially in the cases where the number of samples was relatively

low. Hence, to our knowledge, we provide here, for the first time, the JEV

evolutionary rate estimates confirmed by a formal statistical test.

The comparison showed that the evolutionary rate of the JEV common clade

including all five genotypes GI-GV obtained in our study (95% HPD, 1.67×10^{-4}–

3.14×10^{-4} s/s/y) is significantly lower than values reported by Mohammed et al.

(2011) (95% HPD, 3.49×10^{-4}–5.30×10^{-4}) that can be explained by the smaller

sample size (n=35). In contrast, our estimates are significantly higher than the JEV

substitution rates provided by Pan et al. (2011) (95% HPD, 6.04×10^{-5}–1.69×10^{-4})

and Gao et al. (2015) (95% HPD, 4.37×10^{-5}–1.56×10^{-4}) who analyzed 98 (GI-GIV)

and 100 (GI-GV) sequences, respectively. These disparities may be accounted for
by the lack of temporal signal in the data sets of about 100 sequences. In the case of
low informative data, the Bayesian analysis turns out to be very sensitive to a prior
choice. Therefore, an inferred substitution rate is largely rested on, for example, a
continuous-time Markov chain (CTMC) rate reference prior — gamma probability
distribution with a shape parameter $\alpha=0.5$ and rate parameter $\beta=T$, where T is tree
length (Ferreira and Suchard, 2010) — which is, in turn, ultimately conditioned by
a coalescent population size prior (i.e., the more population size, the longer tree).

In one of the last studies, Xu et al. (2023) reported a JEV GI-GV substitution
rate (95% HPD, 1.62×10^{-4}–3.97×10^{-4}) comparable with our estimates. Notably, the
sample size (115), clock model (UCLD) and tree prior (Bayesian Skyline) applied
in the work of Xu et al. (2023) were similar to two previously discussed studies with
significantly lower JEV substitution rates (Pan et al., 2011; Gao et al., 2015). The
possible explanation is that Xu et al. (2023) provide new JEV nucleotide sequences,
however without testing at least the measure of clock-like evolution by RTT, the
question about the true factor of distorted inference of substitution rate persists
unresolved.

Our estimates of JEV GI evolutionary rate (95% HPD, 3.45×10^{-4}–4.82×10^{-4})
are significantly higher than those calculated for the GI-GV clade (95% HPD,
1.67×10^{-4}–3.14×10^{-4}) (Figure 3). GI rate values inferred in our study are also
statistically different from the literature data, except for the estimates with the widest
95% HPD interval (3.42×10^{-9}–2.47×10^{-4}) provided by Mohammed et al. (2011).
This wide width of the 95% HPD interval is likely due to a small sample size (n=10)
and, as a consequence, the absence of a temporal signal which was not evaluated.
The only study (Li et al., 2023) in which the extent of time-dependency in data was
assessed (RTT) reported statistically higher JEV GI evolutionary rates (95% HPD,
5.03×10^{-4}–6.94×10^{-4}) than our estimates. We speculated on two possible
explanations of the discrepancy — the first one is Li et al., 2023 removed RTT
regression outliers without explaining the reason (e.g., sequencing errors, long
passage history, lab contamination, etc.) that is forcing the data to fit a strict clock

model, which does not assume among-lineage rate variation, instead of using a

proper model (e.g., relaxed clock) to describe the data; the second one is the new

JEV nucleotide sequences generated in the study of Li et al. (2023) and not included

in our work.

Among all three data sets analyzed (Table 1), JEV GIII was characterized by

the lowest evolutionary rate (95% HPD, 3.84×10^{-5}–8.63×10^{-5}) (Figure 3). Thus,

JEV GIII evolves on the other order of magnitude than JEV GI and JEV GI-GV. The

very similar estimates (4.28×10^{-5}–1.18×10^{-4}) were reported by Pan et al. (2011) with

a sample size of 53 and a model combination similar to the one we applied in our

work with the exception of a relaxed molecular clock (UCLD). In the study of

Mohammed et al. (2011), the 95% HPD interval was the widest one (7.05×10^{-8}–

6.03×10^{-4}) like in the case of JEV GI and the sample size was probably one of the

main reasons for that (n=22). To date, JEV GIII substitution rate estimates derived

in our study are relatively the most accurate.

To summarize, GI is the fastest-evolving genotype of JEV and, conversely,

GIII is the slowest JEV genotype. Concerning 95% HPD interval width, our JEV

rate estimates are the most or one of the most robust.

3.3.2 Comparing the substitution rate of JEV (GI-GV clade) and other

mosquito-borne flaviviruses

To provide a better understanding of the JEV evolutionary rate, we compared

the rate of JEV GI-GV clade obtained with rates of JEV most closely related species

and the other well-studied mosquito-borne flaviviruses (Figure 4).

For the comparison of the JEV substitution rate, we considered the data only

if 95% HPD intervals were provided. Also, the estimates reported with mistakes or

obvious typos were excluded from the comparison. We mainly focused on the

substitution rates derived from the analysis of as many virus genotypes as possible

(i.e., the genetic diversity of viruses was meant to be represented to its greatest

extent).

Figure 4. Comparison of the Japanese encephalitis virus (JEV) substitution rate inferred in our study using 325 complete open reading frame nucleotide sequences of the genotype I–V clade (GI-GV) and the substitution rates for the other mosquito-borne flaviviruses according to the literature: USUV—Usutu virus (Engel et al., 2016; Zecchin et al., 2021), MVEV—Murray Valley encephalitis virus (Williams et al., 2015), WNV—West Nile virus (May et al., 2011; McMullen et al., 2013; Srihi et al., 2021), YFV—yellow fever virus (Delatorre et al., 2019; Moreira Salles et al., 2022), ZIKV—Zika virus (Sun et al., 2017; Liu et al., 2019), DENV—dengue virus (Stica et al., 2022). The vertical bars represent 95% highest posterior density intervals, circles—median or mean values if available. The X-axis labels are references to appropriate studies. The methods on temporal signal evaluation applied are indicated above the X-axis: BETS—Bayesian evaluation of temporal signal, RTT—root-to-tips regression, PT—permutation test, *—temporal signal was not assessed.

Unlike the literature data on JEV, the most of MBFV rate estimates were supported by the RTT regressions, and, furthermore, a permutation or randomization test was applied in one study (Srihi et al., 2021). This makes the comparison much more reliable, but, for some virus species, reported rates are statistically different between studies.

First, we compared our JEV substitution rate estimates with the rates of JEV most closely related species (Moureau et al., 2015) — Usutu virus (USUV), Murray Valley encephalitis virus (MVEV), West Nile virus (WNV) — for which the sample size allows the evolutionary rate of the virus to be reliably evaluated. Second, we

included in the comparison the most widely understood members of MBFVs —

yellow fever virus (YFV), Zika virus (ZIKV), and dengue virus (DENV).

As can be seen in Figure 4, JEV is one of the slowest changing MBFVs represented in our analysis (95% HPD, 1.67×10^{-4}–3.14×10^{-4}). The overlap of 95% HPD intervals is observed in the case of two assessments of WNV (2.24×10^{-4}–5.34×10^{-4} according to McMullen et al. (2013) and 2.50×10^{-4}–4.47×10^{-4} for WNV lineage 2 reported by Srihi et al. (2021)). There is the slight overlap with the rate values of the USUV European lineage (95% HPD, 3.04×10^{-4}–3.77×10^{-4}). More distant MBFV members evolve faster, especially ZIKV (Sun et al., 2017; Liu et al., 2019) and DENV (Stica et al., 2022), the substitution rates of which are near the threshold of 10^{-3} and 10^{-2} orders of magnitude or even above it.

3.4 Divergence times

The evaluation of JEV divergence times was performed using 325 complete ORF nucleotide sequences (Figure 5). The results of the analysis in BEAST suggest that JEV emerged 1234 years ago (95% HPD, 407–2333 years).

Figure 5. Maximum clade credibility tree of JEV summarized from a posterior tree distribution produced by an MCMC run in BEAST. The numbers nearby the main nodes indicate divergence times before the most recent sample collected on February 20, 2022, with 95% highest posterior density (HPD) intervals within parantheses. The horizontal dark green bars depict 95% HPD of the divergence times of JEV genotypes.

340   Despite the relatively large number of complete ORF sequences used in our
341 study, the uncertainty in the assessment of the evolutionary timescale of the entire
342 tree spans a large time interval (95% HPD of root height, 407–2333 years). A little
343 more accurate estimates of time to MRCA of JEV GI-GV (95% HPD, 463–2100
344 years) were reported by Xu et al. (2023) with a sample size of 115 complete genome
345 sequences. In the other studies with smaller sample sizes, 95% HPD intervals for
346 tree roots were much wider: Gao et al. (2015) conclude that the MRCA of JEV GI-
347 GV is estimated to have occurred 3255 years ago (95% HPD, 978–6125 years); Pan
348 et al. (2011) studied JEV GI-GIV defined the MRCA age of the entire clade as 1695
349 years ago (95% HPD, 548–3153 years). More recent JEV emergence was described
350 by Mohammed et al. (2011) (95% HPD, 374–500 years ago), but the sample size
351 was the smallest one (n=35) which implies weak temporal structure in the data and
352 is also reflected in the substitution rate estimates (Figure 3).

353   Concerning the emergence of JEV genotypes, 95% HPD intervals of time to
354 MRCA of all JEV genotypes are overlapped (Figure 5). Therefore, we are unable to
355 reliably establish the chronology of the occurrence of genotypes. Besides, only the
356 GI and GIII are representative in terms of sample size, and, as a consequence, the
357 estimates of the other genotype ages (GII, GIV, GV) are unreliable. Generally,
358 considering the bounds of 95 HPDs obtained, JEV genotypes occurred 77–540 years
359 ago.

360   The yielded ages of JEV subtypes are consistent with the previous assessments
361 (Xu et al., 2023). However, the "minor" JEV genotypes, such as GII, GIV, GV, were
362 mainly scrutinized in the studies of E gene data sets. The estimates of time to MRCA
363 based on E gene sequences are more accurate (Schuh et al., 2013; Sikazwe et al.,
364 2022), whereas the substitution rate of E gene is significantly higher (95% HPD,
365 3.92×10^{-4}–6.52×10^{-4}) and MRCA ages are younger (e.g., the MRCA of JEV GI-GV,
366 95% HPD 216–921 years) (Schuh et al., 2013). "Rejuvenation" of the tree root along
367 with (or possibly due to) an increase in the evolutionary rate indicates the impact of
368 the priors on the inference of the evolutionary timescale. In other words, even if the
369 E gene of JEV evolved faster than the entire ORF gene, then the dates of the

emergence of JEV MRCA should not significantly differ between the analyses of
two loci in the case of the obligatory presence of temporal signal in both data sets.
In the cited works dedicated to the E gene analysis, only Sikazwe et al. (2022)
performed the RTT test, however, as a result of a test, JEV GV was removed from
the analysis as an outlier without a justified reason. Moreover, the R^2 and the
regression visualization were not reported as well as the substitution rate of the
common data set (JEV GI-GIV). Thus, the problem of the fitness of the E gene for
JEV date-stamped phylogeny reconstruction should be addressed in future studies.

Conclusions

1.   All complete ORF data sets analyzed (JEV GI-GV, GI, GIII) yielded
strong temporal signal in BETS and can be used for the reconstructions of dated
phylogenies;

2.   JEV represented by the clade comprising all five genotypes (GI-GV)
has the lowest substitution rate in comparison to its most closely related species and
other well-studied members of mosquito-borne flaviviruses: 2.41×10^{-4} s/s/y (95%
HPD, 1.67×10^{-4}–3.14×10^{-4} s/s/y) or approximately 2.5 nucleotide substitutions per
ORF (10,296 nucleotides) per year. To our knowledge, it is the first reliable
assessment of the JEV substitution rate confirmed by the formal Bayesian test of
temporal signal;

3.   JEV GIII is characterized by the lowest substitution rate (6.17×10^{-5}
s/s/y (95% HPD, 3.84×10^{-4}–8.63×10^{-4} s/s/y) or approximately 0.6 nucleotide
substitutions per ORF per year) in comparison with GI and the common JEV clade.
It is the most accurate estimate we found in the literature;

4.   Time to JEV MRCA (GI-GV clade) has the mean value of 1234 years
(95% HPD, 407–2333 years);

5.   The mean values and appropriate 95% HPD intervals of divergence of
each JEV genotype are comparable and close. Based on the lower and upper bounds
of HPDs, JEV genotypes occurred in the period from 77 to 540 years ago.

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JEV GI-GV

R²=0.04

JEV GI

R²=0.79

JEV GIII

R²=0.12