The influence of roads on the fine-scale population genetic structure of the
dengue vector Aedes aegypti (Linnaeus)

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Abstract

Dengue is endemic in tropical and subtropical countries and is transmitted mainly by Aedes aegypti. Mosquito movement can be affected by human-made structures such as roads that can act as a barrier. Roads can influence the population genetic structure of Ae. aegypti. We investigated the genetic structure and gene flow of Ae. aegypti as influenced by a primary road, España Boulevard (EB) with 2000-meter-long stretch and 24-meters-wide in a very fine spatial scale. We hypothesized that Ae. aegypti populations separated by EB will be different due to the limited gene flow as caused by the barrier effect of the road. A total of 376 adults and larval Ae. aegypti were collected from June to September 2017 in 21 sites across EB. North (N1-N11) and South (S1-S10) comprised of 211 and 165 individuals, respectively. All mosquitoes were genotyped at 11 microsatellite loci. AMOVA $F_{ST}$ indicated significant genetic differentiation across the road. The constructed UPGMA dendrogram found 4 genetic groups revealing the clear separation between North and South sites across the road. On the other hand, Bayesian cluster analysis showed four genetic clusters ($K = 4$) wherein each individual samples have no distinct genetic cluster thus genetic admixture. Moreover, significant positive spatial autocorrelation was observed at 100 - 200 m distance class, suggesting dispersal potential of the adult mosquitoes within a 200 m spatial scale. Our results suggest that human-made landscape features such as primary roads are potential barriers to mosquito movement thereby limiting its gene flow across the road. This information is valuable in designing an effective mosquito control program in a very fine spatial scale.

Author Summary

Dengue, a mosquito-borne viral infection is a serious health problem in tropical and subtropical countries such as Philippines. Most dengue prevention programs aim to eradicate its mosquito vector, Aedes aegypti. A successful population control program is reliant in understanding
the mosquito behavior and ecology including how human-made structures such as roads influence its expansion and movement. Previous studies have discovered the barrier effect of roads in the movement of mosquitoes. In this study, we examined the influence of roads in the population genetic structure of *Ae. aegypti* in a fine spatial scale using 11 microsatellite markers. We found significant genetic differentiation of mosquito populations across the road. Our results suggest limited gene flow across the road and supports our hypothesis that roads are potential barriers to mosquito dispersal. This information can be used in designing an effective mosquito population control zones in perceived barrier to mosquito dispersal such as roads.

**Introduction**

Dengue is an infectious disease transmitted by *Aedes* mosquitoes. The main dengue vector, *Aedes aegypti* is best adapted to urban environments. Dengue prevention programs are usually focus in eliminating mosquito larval habitats and adult mosquito populations [1 – 2]. Successful dengue control programs are dependent upon understanding the population genetic structure of *Ae. aegypti* including how human-made structures such as roads influence its dispersal. For example, estimating gene flow and barriers to dispersal such as roads can help in predicting the spread of insecticide resistance genes in *Ae. aegypti* populations [3] and *Wolbachia*-infected mosquito populations release [4-5].

Most population genetic studies of *Ae. aegypti* were usually described at a country spatial scale [6-8]. In contrast, fine spatial scale genetic analysis (e.g scale of several households or city block) though challenging is still feasible as evidenced in several studies that revealed significant genetic differentiation of *Ae. aegypti* at fine spatial scales. For example, significant genetic differentiation was found at spatial scales of 5 km to 2,000 km using ND4 mitochondrial gene [9] and microsatellite markers [10-12]. Recently, [13] revealed the low genetic differentiation and high
gene flow among *Ae. aegypti* populations in Metropolitan Manila, Philippines, which suggest the influence of passive and active dispersals of the mosquitoes to population genetic structure. Dispersal ability is a determinant factor of population genetic structure and the genetic effects of habitat fragmentation at fine spatial scales [14]. Population genetics can be utilized in estimating dispersal ability through spatial autocorrelation analysis. For example, limited spatial ranges of significant spatial autocorrelations of up to 1 km suggested the active dispersal capability of *Ae. aegypti* at microgeographic areas of eastern Thailand [12] and in Metro Manila, Philippines [13].

Although the genetic effect of roads in the dispersal pattern of *Ae. aegypti* in a fine spatial scale is very limited, the influence of roads was often studied through conventional mark-release-recapture method (MRR). This method is labor-intensive and the rearing and marking procedure can affect the mosquito fitness and movement in the field [15]. Studies in MRR of *Ae. aegypti* demonstrated that the dispersal of this mosquito vector could be influenced by the type of road [16-17]. *Ae. aegypti* prefer crossing smaller and quieter roads as compared to larger and busy roads [16-17]. To date, the only study that investigated the effect of road such as highway on *Ae. aegypti* using genetic approach was done by [18]. Using larval samples, [18] found significant genetic differentiation across a 900-meter-long stretch and 120-meters-wide highway. Our study, provided an extensive sampling of adult mosquitoes and larval survey in a primary road, España Boulevard (EB). As compared to highways, primary roads are characterized with presence of several households, buildings and minor residential roads across that could serve as a pathway for public and private transportation to traverse. We expect that passive method of mosquito dispersal in primary roads are possible than in highways because of the presence of minor roads that can aid the movement across.

Furthermore, genetic analysis may also reveal co-occurrences of multiple genetic clusters at fine-spatial scale and their genetic admixture in mosquito individuals. For example, previous
studies of *Ae. aegypti* revealed sympatric numerous genetic clusters (K = 3 to K = 16) from 30 km up to 2,000 km [10-11,13,19], which might be due to the divergence from a single ancestry resulting into multiple genetic clusters over time and the random distribution of *Ae. aegypti* populations from nearby cities, regions or country. A recent study in the region of Metropolitan Manila, Philippines [13], discovered the probable number of genetic clusters of K = 4 in a fine spatial scale in Metro Manila, Philippines and genetic admixture in *Ae. aegypti* individuals. In this study, we narrow down the spatial scale up to 2 km to test if we could still observe multiple genetic clusters. This information is important prior to analysis of population genetic structure and gene flow because it can give background information on how the variety of genes are co-existing in a limited spatial scale.

Here, we studied the influence of road on the population genetic structure and gene flow of *Ae. aegypti* using 11 microsatellite loci, to analyze the genetic relatedness among the mosquito populations and to determine the number of genetic clusters in a very fine spatial scale (2,000 m x 24 m EB road). We hypothesize that *Ae. aegypti* populations across the España Boulevard (EB) with 2000-meter-long stretch and 24-meters-wide may have differences in its population genetic structure because EB act as a potential barrier in the mosquito movement. Previous study of [18], indicated a barrier effect of a highway to *Ae. aegypti* dispersal using all larval samples from water holding containers. Highways have continuous traffic without the presence of secondary roads that can be utilized as pathway for mosquito movement. To provide more information about the effect of roads, we collected adult mosquitoes and representative larval samples from each water holding containers across a primary road with presence of several smaller roads that can potentially carry mosquito migrants across. The sampling method of collection of both adult *Ae. aegypti* and larvae though labor intensive is more informative as compared to only larval samples.

**Methods**
Study site

The study selected a certain area in City of Manila, Philippines. This area consists of two traversing roads, a primary road España Boulevard (EB) and a secondary road AH Lacson Avenue (LA). EB is the selected study area and is divided into North and South sides. It is a primary road based on Philippine geographic information system (PhilGIS). It is located within a highly urbanized area in Metro Manila consisting of commercial, residential, and industrial infrastructures and it connects two cities: Manila City and Quezon City. EB as a primary road have 44 intersections connecting two or more roads across the north and south. The mean width of the road (EB) sampled is 24.27 meters and its length is 2,000 meters.

Collection, sampling and identification

We used a two stage cluster systematic sampling design to randomly select households for collecting mosquitoes. We used OpenEPI software [20] to calculate the target sample size. Equal allocation and a design effect of 3 [21,22] was used to calculate the target sample size. The estimate of \( p \) used in the calculation was 0.23 according to the study of [23]. The alpha level was set at 95% (\( \alpha = 1.96 \)). The maximum tolerable error was equal to 10%. We computed the sample size on OpenEpi online software (https://www.openepi.com/Menu/OE_Menu.htm) [20]. An additional 15% allowance was added to provide a buffer, including refusal to participate, yielding a target sample size of 236 households per stratum (North and South; \( n = 472 \)). The sampling unit of our study was the household defined as one unit of accommodation [24]. Each site (n=21) is comprised of 1 to 3 smallest administrative division within the city (barangay; \( n = 19 \)) and households (\( n = 10 \) to 20 per site) were selected based on their voluntary informed consent for mosquito collection and larval survey. Each site is defined as the population of this study. The geographic coordinates of the...
sampled households were recorded. In order to obtain a single geographic coordinate for each sampling site, we calculated the geographic midpoint [25] of sampled households.

Collections of adult and larval mosquitoes were performed during the rainy season and high dengue cases, from June to September 2017. Adult mosquitoes were collected by commercially available mosquito UV light trap (MosquitoTrap, Jocanima Corporation, Las Piñas, Philippines). The traps were set for 48 hours inside each household. We also surveyed potential water breeding sites on the household premises and collected larvae when present. Mosquitoes and larvae collected were identified morphologically to species level using the keys for adult and immature mosquitoes of [26]. All collected samples were preserved in RNALater™ and stored at -20°C prior to nucleic acid extraction.

We analyzed 211 Ae. aegypti from 106 households in the North area and 165 Ae. aegypti collected from 104 households in the South area of the road (S1 Table). The number of Ae. aegypti collected per site ranged from 11 to 33 individuals.

**DNA extraction, PCR amplification and microsatellite genotyping**

Simultaneous DNA and RNA extraction in individual mosquito adult and larval samples \((n=376)\) was performed using the Qiagen® AllPrep DNA/RNA micro kit following the manufacturer’s protocol with some modifications. In this study, the extracted DNA was only used while the extracted RNA were kept in – 80°C for future investigations (e.g. transcriptome analysis). DNA quality was checked in NanoDrop™ 2000 Spectrophotometer (Thermo Scientific). We used the same 11 microsatellite primer and PCR protocol that have been used in an earlier study by [13]. All PCR amplifications were performed with 1µl of genomic DNA in a final volume of 10 µl. We performed 4 sets of multiplex PCRs with 3 pairs of loci grouped together (S2 Table). All PCR amplifications were composed of the following: 10x Ex Taq buffer, 25mM MgCl₂, 2.5mM dNTP, 5% dimethyl sulfoxide
(DMSO), 10µm of fluorescently labelled forward primer, 10µm of reverse primer, and 5 units/µl of 
Takara Ex Taq™ (Takara Bio Inc.). Each multiplex PCR amplification was conducted as follows: an 
initial denaturation of 94°C, denaturation at 94°C, annealing varies in each set of multiplex PCR (57°C 
to 62°C), extension at 72°C, and final extension of 72°C. PCR products were analyzed in 3% agarose 
gel electrophoresis stained with Midori Green Advance DNA stain.

Multiplex PCR products were diluted in 1:3 water and pooled into final volume of 16 µl.

Samples were prepared prior to fragment analysis with 1 µl of pooled PCR product added with 0.5 
µl GeneScan 600 LIZ dye standard and 10 µl HI-DI Formamide. Fragment analysis was performed 
using SeqStudio Genetic Analyzer (Applied Biosystems). We used PeakScanner (ThermoFisher 
Scientific) to identify peak and fragment size and Microsatellite Analysis app (ThermoFisher 
Scientific) for genotyping. The dataset was made publicly available and deposited at vectorbase.org 
with the population biology project ID_____.

Data analysis

Allele scores were checked for genotyping errors and for the presence of null alleles using 
Microchecker [27]. The observed heterozygosity (H_o), expected heterozygosity (H_e), mean number 
of alleles, mean number of effective alleles, allelic richness, mean number of allele frequency and 
mean number of private alleles were computed in GenAlEx version 6.51b2 [28]. We calculated the 
Inbreeding coefficient (F_is) for all loci across populations following [29] and tested statistical 
development from Hardy-Weinberg equilibrium (HWE) using Genepop web version [30]. The markov 
chain parameters were set at 10,000 dememorizations, 100 batches and 5,000 iterations for testing 
development from HWE.

To test the statistical significance of genetic variations among groups (North and South of 
EB), among sites (N1 – N11 and S1 – S10) within groups, and within sites (N1 – N11; S1 – S10), we
computed the Analysis of Molecular Variance (AMOVA) using Arlequin version 3.5.2.2 [31] with 10,000 permutations. We assessed the degree of genetic differentiation between the 21 sampling sites by calculation of the pairwise $F_{ST}$ values in Arlequin. In order to determine if the mean values of the pairwise $F_{ST}$ within groups (North versus South) and between groups are significantly different from each other, we performed Mann-Whitney U-test.

Dendrograms among the sites were constructed using the genetic distance matrix (pairwise $F_{ST}$ values) generated from Arlequin software. We employed the Unweighted Pair Group Method with the arithmetic mean (UPGMA) method using the APE package [32] and R program [33]. To determine the optimal number of groups in the dendrogram, we used the pseudo-$t^2$ index from the package NbClust [34] of R program.

To infer the individual assignment of *Ae. aegypti* to genetic clusters whose members share similar genetic characters, we used the Bayesian clustering algorithm in STRUCTURE version 2.3.4 software [35]. We used the same parameter set as in [13] testing for 1-20 presumed genetic clusters (K) with 20 iterations per K, a burn-in period of 200,000 steps and 600,000 Markov Chain Monte-Carlo (MCMC) replications using an admixture model with correlated allele frequencies. The best estimate of K was calculated with the ad-hoc statistic $\Delta K$ as described by [36] using Structure Harvester Web version 0.6.94 (http://taylor0.biology.ucla.edu/structureHarvester/#) [37]. We visualized the final barplots using the R package pophelper [38] as implemented in R program.

The test for Isolation by distance was performed using a Mantel’s test in GenAlex to determine if geographical distance influence the genetic differentiation. The pairwise genetic distance ($F_{ST}$) was compared to the geographical distance (km) among the sites. To obtain the geographic distances between sites we used the geographic midpoint of the sampled households per site calculated based on the coordinates (latitude and longitude) of the households. All Mantel
tests were assessed for the significance of the correlation using permutation tests (9999 permutations).

To further evaluate whether genetic variation was correlated with geographic distance, we performed a spatial autocorrelation analysis using GenAlEx [28]. We computed the autocorrelation coefficient \( r \) from the geographic distance as described above and the genetic distance (pairwise \( F_{ST} \) values). This measure determines the genetic similarity between the 21 sites within an identified geographic distance class. We identified the suitable distance class based on the observed distribution of pairwise geographic distance between sites. We used 14 distance classes at 0.10 km interval.

**Results**

**Genetic Diversity and Differentiation**

The mean number of alleles (MNa) per sampling site ranged from 4.55 (S7) to 10.73 (N2) while the mean number of effective alleles (Mne) ranged from 2.86 (S9) to 4.54 (N2) (S1 table). In contrast, the mean number of allele frequency ranged from 3.18 (S10) to 4.91 (N9, S2 and S8) between sites and the mean number of private alleles ranged from 0.00 (N9, S6 and S9) to 0.82 (S3) between sites. All 21 sites displayed non-conformance to Hardy-Weinberg equilibrium (He > Ho) which implies heterozygosity deficiency that can be caused by inbreeding, the expected heterozygosity (He) ranged from 0.55 (S7) to 0.72 (N11).

AMOVA results showed significant genetic differentiation \( (F_{ST} = 0.0336) \) between sampling sites North (N1 – N11) and South (S1 – S10) of the EB road (Table 1). Small but significant estimates among \( (F_{SC}) \) and within \( (F_{CT}) \) 21 sites (N1 – N11; S1 – S10) were observed. Population pairwise \( F_{ST} \) between the combined all northern and all southern sites showed significant genetic differentiation \( (F_{ST} = 0.0321) \). A significant difference between the mean pairwise \( F_{ST} \) within groups (North and
South; mean = 0.0321) and between groups (mean = 0.0337) were found using the Mann-Whitney U-test at P < 0.05. The pairwise $F_{ST}$ among the 21 sites ranged from 0.0029 (N2 and N8) to 0.1085 (N6 and S7). Among these comparisons of pairs of sites, 110 out of 201 (54.73%) pairwise $F_{ST}$ values presented significant genetic differences (S3 Table).

Table 1. Analysis of molecular variance (AMOVA) using a panel of 11 microsatellites

| Variation                        | ss      | vc       | pv       | Fl                        |
|----------------------------------|---------|----------|----------|---------------------------|
| Among North & South              | 18.6330 | 0.0205   | 0.3864   | $F_{ST} = 0.0336^*$        |
| Among sites within North and South| 203.2070| 0.1579   | 2.9737   | $F_{SC} = 0.0299^*$        |
| Within sites                     | 3726.3140| 5.1317  | 96.6399  | $F_{CT} = 0.0039^*$        |

ss = sum of squares; vc = variance components; pv = percentage variation; F-statistics for each hierarchy; $F_{ST}$ = among groups; $F_{SC}$ = among populations within groups; $F_{CT}$ = within populations; *P < 0.05

The number of groups identified on the UPGMA Dendrogram was four based on the cindex index. The groupings (Fig 2) revealed the clear separation of North (Group 1 – N1 to N3; N5; N7 to N10) and South (Group 3 – S4 to S10) of EB. On the other hand, genetic similarity between some sites from the North and South of EB were shown for groups 2 (N4, N11 and S1, S3) and 4 (S2 and N6).
Fig 1. An unweighted pair group method with the arithmetic mean (UPGMA) dendrogram and map of the 21 sites where the *Ae. aegypti* were collected in households across EB highlighted in red and in black is LA, an important East-west connection. For more detailed information, please see S1 Table.
A. A dendrogram based on the unweighted pair group method with the arithmetic mean (UPGMA) using the pairwise genetic distance among the 21 sites from the north and south of EB road. B. Map showing the distribution of the genetic groups as observed in the UPGMA cluster dendrogram (Fig 2 A). The color of the triangles corresponds to the results of the dendrogram.

**Genetic Structure**

In STRUCTURE analysis, the most probable number of genetically differentiated clusters across the mosquito populations was $K = 4$ (S1 Fig). The barplot (Fig 3) displays the distribution of the assumed genetic clusters of each *Ae. aegypti* in the North (N1 to N11) and South (S1 to S10) of the road. The barplot suggests admixture of the genetic clusters across all the mosquito individuals.

![Barplot of genetic structure](image)

**Fig 3.** STRUCTURE bar plot displaying the assignment probabilities of each genotyped *Ae. aegypti* individual grouping into 4 clusters. The x-axis shows the individual and collection sites from the North (N1 to N11) and South (S1 to S10).

**Isolation by Distance and Spatial Autocorrelation**

The Mantel test analysis based on the pairwise genetic distances ($F_{ST}$) and the geographic distances between all populations was not significantly correlated ($r = 0.092$, p-value = 0.138) (S2 Fig) thus indicating no isolation by distance. On the other hand, the spatial autocorrelation showed a positive and significant correlation between genetic and geographic distance from 0.10 - 0.20 km. (Fig 4).
Fig 4. The figure shows the results of the spatial autocorrelation analysis as the influence of geographic distance on the genetic distance. The correlation coefficient $r$ is plotted against geographic distance (blue line) and the upper and lower 95% confidence intervals are indicated as red dotted lines. The $r$-values above and below the dashed lines indicate spatial structure. The error bars are derived from 10,000 permutations of individuals FST values within each distance class.

Discussion

Our findings support the hypothesis that human-made structure such as roads can create a barrier to dispersal of *Ae. aegypti* on a very fine spatial scale. The significant $F_{ST}$ estimate (0.0336) across the EB revealed genetic differentiation on both sides (North and South) of EB as supported by the significant population pairwise $F_{ST}$ ($F_{ST} = 0.0321$) between the combined all North sites and all South sites. Our result is consistent with a study conducted in a 900-meter long and 120-meter wide road in West Indies, which found a small but significant $F_{ST}$ value (0.011 to 0.021) across the road using 9 microsatellite loci and 2 SNP markers [18].

The results of the cluster dendrogram showed the clear separation of North sites and South sites that further supported our hypothesis of barrier effect of roads in mosquito movement. Group 1 (Fig 1b, blue line) illustrates the clustering of the genetically similar *Ae. aegypti* populations from
the north side of the EB while Group 3 (Fig 1b, green line) showed the clustering of populations from the south sites. The EB road may separate the two areas potentially limiting the migration between mosquito populations thereby resulting in the formation of genetic groups as seen in the results of the cluster dendrogram. For example, [13] explained that some cities in their study area were marginally separated based on their cluster dendrogram analysis by these landscape features such as highway and that the alleles could be generally preserved and the allelic combination maybe distinct from other mosquito population overtime due to limited gene flow.

Our results of the cluster dendrogram and pairwise $F_{ST}$ are interesting because the mean width of the road is around 24.27 m and given the dispersal capability of *Ae. aegypti* ranging from 100 m to 800 m [39 – 43], we should not see distinct groupings and no limited adult dispersal unless the mosquitoes were not able to successfully cross EB road. In addition, positive spatial autocorrelation at the range of 100 m to 200 m implies the active dispersal capability of *Ae. aegypti*. Despite the fact that the distance of the EB road is within the dispersal estimates of *Ae. aegypti*, possibly the inadequate cover and shade from trees and vegetation across the EB road made it unsuitable for the mosquitoes to traverse across as previously explained by [18]. The degree of shade has strong interrelation with the presence of *Ae. aegypti* in mosquito and larval surveys [44]. Mosquitoes tend to disperse in areas with numerous water breeding containers and trees that can provide heavy shading while busy roads seem to inhibit the mosquito movement [17]. The lack of available oviposition sites and suitable blood meal hosts [18] are possibly some of the factors that have prevented *Ae. aegypti* movement across EB road.

Interestingly, Group 2 (N4, N11, S1 and S3) (Fig 1, orange line) and Group 4 (S2 and N6) (Fig 1, red line) of the dendrogram displayed clustering of sites from the north and south. The co-presence of South sites (S1 and S3) and North sites (N4 and N11) in Group 2 could be attributed that a part of the EB is not a potential barrier and migration of mosquitoes could be possible. For example,
the presence of secondary roads (e.g. LA road and the intersections) probably facilitated the passive
dispersal of mosquitoes between sites N4, N11 and sites S1, S3 for example through increased
human-mediated dispersal of *Ae. aegypti*. Meanwhile, Group 4 showed that S2 and N6 sites mixed
from the west end of the road with presence of crossing minor roads such as residential, pedestrian
lane and intersections that can act as a route of passive mosquito dispersal. Minor roads could be a
possible route for mosquitoes to traverse the road. For example, small pedestrian lanes provide
passive mosquito dispersal by humans such as when mosquitoes are accidentally transported by
land vehicles that can be over the flight distance capability of a typical *Ae. aegypti* [43,45-46]. Future
investigations are necessary to validate the kinds of conclusions that can be drawn from the effects
of ecological factors mentioned previously. Alternative mechanisms that drive the mixed structure
in Groups 2 and 4 might be because of the erroneous estimation of allele frequencies of populations
due to small sample sizes per populations and common source population/s between the sites in
group 2 or group 4 allow to share same alleles across the road.

We detected multiple genetic clusters (K = 4) in a very fine spatial area by Bayesian analysis,
which was concordant with the results (K = 3 to 4) from a previous study of *Ae. aegypti* among 11
sites not more than 30 km apart in Sao Paulo, Brazil [11]. Previous population genetic studies of *Ae.
aegypti* from Philippines also displayed multiple genetic clusters (K= 2 to K = 6) in fine-spatial scale
using microsatellite markers [13, 47-48]. The different ancestry populations of our samples may be
the reason of the co-occurrence of the four genetic clusters in the small area rather than the limited
gene flow in the study area. Despite the occurrence of multiple genetic clusters (K = 4), our results
revealed genetic admixture thus no distinct genetic cluster observed. The genetic admixture might
indicate that the individual *Ae. aegypti* from these sites could potentially share alleles possibly due
to the several mosquito invasions from neighboring cities surrounding the study area as observed
in Philippines [13], China [49] and in the USA [50].
Our sampling strategy of collecting adult and representative larval samples from surveyed water breeding containers though challenging is more informative in the analysis of mosquito populations in a very fine spatial scale. We assume that our sampling strategy of adult mosquitoes’ collection are more likely to represent the sampling site as compared to collecting or sampling eggs or larvae from the same water container [51]. Adult mosquito sampling within household can increase the chance of getting higher genetic variability as compared to only larval sampling. In contrast to [18], our study sampled adult mosquitoes (n = 359) and single larva from each surveyed water breeding container (n= 17) to minimize the possibility of sampling family members as seen from the previous studies of [52-55]. The collection of one larvae per water breeding container throughout the sampling site minimizes the probability of larvae from the same progeny [56].

Overall, the findings of this study displayed strong evidence of limited gene flow across the highway causing habitat fragmentation of the mosquito populations from the north and south of EB road. The results suggest that human-made structures such as primary road are potential barriers to mosquito dispersal limiting its movement across the road. Understanding the dispersal pattern of Ae. aegypti in a very fine-spatial scale can give insights in predicting the spread of dengue virus infection. This information can also be used in the design of successful vector control strategies such as mosquito elimination programs in a very fine spatial scale. For example, we can use the information on the blocking potential of roads in Wolbachia-infected mosquito release programs. Road can be used as a unit of release as compared to city-wide mosquito release. Local elimination of Ae. aegypti can also be achieved by assigning control zones along roads that can potentially block the mosquito movement. Road blocking information can be used during dengue outbreaks wherein vector control agencies can determine high risk areas in the control zones. Knowledge on the effect of human-made structures such as roads in mosquito dispersal can greatly improve the implementation of a successful mosquito control programs [18].
Acknowledgements

We are grateful to Johanna Beulah Sornillo of Research Institute for Tropical Medicine, Philippines for her technical expertise and assistance in the statistical sampling design of our study. We would also like to thank Dr. Crystal Amiel Estrada of the University of the Philippines – Manila for providing her valuable comments regarding the conceptualization of the research study. We would like to thank Katherine Viacrusis and Tatsuya Inukai for their technical assistance during the mosquito collection, Dewi Gustari for her assistance in the PCR experiments and Micanaldo Francisco for his technical assistance in constructing the sampling maps of this study. The authors are also thankful to Dr. Mary Jane Flores for her suggestions about the sampling strategy. We are also thankful to Dr. Maribe Gamboa, Dr. Michael Monaghan and Joeselle Serrana for their valuable suggestions on the population genetic analyses.

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57. Supporting Information
S1 Fig. Identified optimal number of clusters (K) as calculated in the Structure Harvester program.

S2 Fig. Mantel test on the relationship between genetic distance (pairwise $F_{ST}$) and geographic distance between to sites in km of the 21 sites of *A. aegypti*

S1 Table Summary of variation at 11 microsatellites, north and south of the EB road

S2 Table List and characteristics of microsatellite markers used in genotyping

S3 Table Pairwise $F_{ST}$ values in bold indicates significant p value $<0.05$. Sites are labelled as N1 – N11 (North) and S1 – S10 (South)