Genetic Diversity and Population Structure in *Elaeidobius kamerunicus* (Coleoptera: Curculionidae) Inferred from mtDNA COI and Microsatellite Markers

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**ABSTRACT**

*Elaeidobius kamerunicus* Faust (Coleoptera: Curculionidae) is widely known as the main pollinator of oil palm. Although economically important, the population genetics of this pollinator have remained relatively unexplored. This study explores population genetic structure and contemporary gene flow in *E. kamerunicus* in Indonesian provinces and attempts to place observed patterns within the broader geographical context. Based on combined data from mtDNA cytochrome oxidase subunit I (COI) sequences and 11 SSR loci obtained from 32 individuals located in 11 provinces and 16 sites, overall genetic structuring was low. The study suggested a close relationship of individuals between close provinces but some of data showed that individuals from provinces with far geographical also shared some close relationship. The expansion history of this species, including likely human-mediated dispersal, may have played a role in shaping the observed weak structure detected in this study. The information resulted from this analysis of gene flow and population structure has broad implications for monitoring and management of this pollinator, especially in oil palm plantations in Indonesia. Future studies should concentrate efforts on sampling of African and South-East Asian populations, which would enable better inferences of the ancestral location of *E. kamerunicus* and its invasion history into and throughout world.

**Keywords**: genetic diversity, geographical, mtDNA, population, SSR

1. **INTRODUCTION**

Insect pollinator behaviour in foraging on flowers influenced by intrinsic (e.g., thermo-regulation abilities) and extrinsic (e.g., ambient temperature or light intensity) factors \(^1\)\(^–\)\(^5\). *Elaeis guineensis* Jacq. produces unisexual male and female inflorescences \(^6\), which are produced in alternate cycles. Thus, oil palm reproduction requires cross-pollination carried out by several agents, one of which is an insect \(^7\). Recently, the beetles in the Curculionidae and Nitidulidae (Coleoptera) families are the most important insect pollinators of oil palms. These pollinators have a mutualistic symbiosis with oil palm; the post anthesis male inflorescences serve as brood sites, and the female inflorescences are pollinated by deceit, as explained in the following paragraph.

The underlying pollination process is mediated by the release of inflorescence odour. Female inflorescences of *E. guineensis* emit an anise-like fragrance similar to the fragrance emitted by the male inflorescences at anthesis to attract insects \(^8\)\(^–\)\(^10\). Pollination by insect is critical for an economically sustainable oil palm industry because it increases fruit set and consequently oil yields \(^11\), \(^12\). The oil palm industry is dependent on the artificial reinforcement of pollination, using human-assisted pollination or the introduction of insects in areas in which they are absent \(^13\). Thus, the study of the pollination system of *Elaeis* spp and the interactions with insect pollinators of these palms may help improve pollination in oil palm plantations. The biology of the pollinators of oil palm species, particularly the timing of visits to inflorescences, is not well understood.

To date, most of the published studies on insect activity on oil palm inflorescences have focused on *Elaeidobius* weevils, including *Elaeidobius kamerunicus* Faust and *E. subvittatus* Faust, and on a beetle *Mystrops costaricensis* Gillogly (Coleoptera: Nitidulidae). *E. kamerunicus* and *E. subvittatus* are both reportedly visitors on inflorescences of *E. guineensis* in Africa, their country of origin \(^14,15\) and in regions where these species were artificially introduced, such as Asia and South America \(^16\)\(^–\)\(^21\). Reported that *E. subvittatus* also showed morning activity on *E. oleifera* inflorescences in South America \(^11,20\). However, a short second period of activity at the end of the day was observed for *E. kamerunicus* in West Africa \(^15\) and on Hainan Island, China \(^17\).

Information regarding genetic diversity, genetic structure and gene flow are key issues when developing bio-management strategies of these pollinator species...
especially *E. kamerunicus* [22, 23]. The mitochondrial (mtDNA) cytochrome oxidase subunit I (COI) gene is easily amplified [24], maternally inherited and relatively fast-evolving, which allows derivation of recent female-specific evolutionary histories. In contrast, microsatellites (SSRs) are nuclear, bi-parentally inherited, highly polymorphic and easy to isolate, making them particularly informative in the study of contemporary biological invasions [25, 26]. Thus, these two sets of molecular markers are indispensable tools to study the population genetic structure of insect pollinator, including *E. kamerunicus* species.

Taken together, this study provides foundational data for understanding the population dynamics and genetic structure of *E. kamerunicus*, which will contribute greatly toward monitoring and management of this pollinator, especially in oil palm plantations in Indonesia.

### 1.1. Material and Methods

#### 1.1.1. Sample collection

The *E. kamerunicus* samples were collected from oil palm female inflorescence during its anthesis from 11 provinces and 16 sites in Indonesia (Table 1). All weevils were stored in 100% ethanol at −20 °C prior to DNA extraction.

#### 1.1.2. DNA extraction

Total genomic DNA was extracted using the gSYNC DNA Extraction Kit (Geneaid) following the manufacturer’s instructions. The quality and quantity of the extracted DNA were evaluated using a Nanodrop ND-1000 spectrophotometer (Nano-Drop Technologies Inc.) and by electrophoresis of the sample DNA in 1% agarose gel.

#### 1.1.3. Species identification using mtDNA COI

Cytochrome c oxidase subunit 1 (COI) primers were used to confirm the identification of the insect. A 708-base pair (bp) fragment of COI was amplified using primers COI E.kam F (5'-TTGAGGATTTGAGAATTGACT-3') and COI E.kam R (5'-TTGCTGAGTAAAATATGCCCGT-3'). The COI genes were amplified using a Veriti Thermal Cycler (Applied Biosystems, USA). PCR was performed in 50 μL of a reaction containing each of 1 μL 0.25 mM forward and reverse primer, 25 μL Kapa 2G Fast Master Mix (Kapa Biosystem, USA), 22 μL nuclease free water and 1 μL of mycelium template. An initial denaturation step for 3 min at 95 °C was followed by 35 cycles of denaturation for 15 s at 95 °C, annealing for 10 s at 55 °C and extension for 10 s at 72 °C, with a final extension step of 5 min at 72°C. The PCR products were visualized by agarose gel electrophoresis. Then, it is directly sequenced at First Base, Malaysia by automated DNA sequencing with ABI 3730 XL (Applied Biosystems, USA) using COI E.kam primers especially for *E. kamerunicus*, which has been designed and optimized by Astra Agro Lestari based on GenBank.

DNA sequences were edited and aligned using Geneious Prime (www.geneious.com). Sequence analyses and specimen identification were done by inputting the trimmed sequence both in NCBI’s BLAST tool. Molecular Identification was made based on percent similarity result. An unrooted phylogenetic tree using Randomized Accelerated Maximum Likelihood (RAxML) method [27] was created in Geneious Prime Software. Also, haplotype networks construction was inferred using Minimum Spanning Network (MSN) method Bandelt *et al.* (1999) [28] and was drawn using POPART software [29].

#### 1.1.4. Population study using SSR genotyping

The genotyping stage is carried out to determine the size of the allele formed from each individual and the allele frequency in each population. In this study the genotype used Eleven SSR marker loci (Table 2.) At this stage, amplification was carried out by PCR method in all insect individuals using primers. PCR was performed in 50 μL of a reaction containing each of 1 μL 0.25 mM forward and reverse primer, 25 μL Kapa 3G Fast Master Mix (Kapa Biosystem, USA), 22 μL nuclease free water and 1 μL of mycelium template. An initial denaturation step for 3 min at 95 °C was followed by 35 cycles of denaturation for 20 s at 95 °C, annealing for 10 s at 51 °C and extension for 15 s at 72 °C, with a final extension step of 15 s at 72 °C and temporary stored at 4 °C. The amplified sample was then visualized with the Automated CE System Fragment Analyzer using dsDNA 910 Reagent Kit and Prosize 3.0 software. As much as 2 μL of PCR samples were taken and given 22 μL of 1X dilution buffer. Separating gel used is a polyacrylamides gel that has been given intercalating dye.

Allele scoring data were tabulated using Microsoft Office Excel. CERVUS version 3.0 and GENALEX version 6.3 were used for analysed population genetic parameters. Principal component analyses were done using DARWIN Software version 6.0.013. Clustering analysis was conducted using Discriminant analysis of principal components (DAPC) as mentioned by Jombart *et al.* [30] inside an R package adegenet [31, 32].

Haplotype networks construction was inferred using Minimum Spanning Network (MSN) method [27] and Briva distance calculation [33]. The results then were drawn using an R package poppr [34].

The population structures were determined using the STRUCTURE software version 2.3.4 and STRUCTURE SELECTOR version web [35]. The approach by Puechmaille [36] and Evanno *et al.* [37] were used to select the appropriate K clusters for the studied populations. The Structure population was drawn using CLUMPAK software [38].
1.2. Our Contribution

This paper discusses genetic diversity and population structure of *E. kamerunicus* in Indonesia.

1.3. Paper Structure

The rest of the paper is constructed as follows: section 2, serves description of allelic pattern, phylogenetic approach, and population structure of *E. kamerunicus*, and section 3 concludes the results of this paper.
2. RESULTS AND DISCUSSION

In this study we used Mitochondrial markers (COI gene) and Nuclear Markers (SSR) as tool to elaborate genetic diversity and population structure of E.kamerunicus. We will discuss based on these two different markers.

2.1. Mitochondrial markers

Mitochondrial DNA (mtDNA) analysis is generally assumed to be more powerful than allozyme and nuclear DNA markers for revealing historical gene flow versus current gene flow [39]. The cytochrome oxidase I (COI) and II (COII) genes of the mtDNA are useful for the measurement of genetic variation, haplotype identification, construction of phylogenies and population genetic studies in insects [40, 41].

Recently, region near the 5’ end of the mitochondrial COI gene was used as DNA Barcoding standard for animal kingdom, including insect [42, 43]. DNA barcoding, in a broad sense, is simply the use of short, standardized genomic segments as markers for species recognition. Just as species differ in morphology, ecology, and behaviour, they also differ in their DNA sequences.

In this study we will discuss about phylogenetic approach and network approach also combine the result of both approaches.

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**Table 2** List of SSR marker used in this study

| Locus | Forward (5'→3') | Reverse (5'→3') | Sequence | Motif | Repeat | Length (bp) | Product (bp) |
|-------|----------------|----------------|----------|-------|--------|------------|-------------|
| ssr-12 | AGATGGACGTTCTACATTTG GC | GAACGAAGAA ACTGTGGTGGCC scaffold-25 | TG | 13 | 26 | 113 |
| ssr-13 | GGTCTCCAAAAC AAGCATC CGG scaffold-25 | TTATCTCCCGGA AGCGACTGC | ACC | 5 | 15 | 128 |
| ssr-137 | AAAAACACT ACGGACTGC GC | GGTTCTAAAG GCCATAGTGC | scaffold-547 | TG | 8 | 16 | 132 |
| ssr-159 | GCTCAAGCTTT TCAAATAATC GTCC scaffold-637 | GTGGAGTATAC AGCAGGGGG | TA | 7 | 14 | 110 |
| ssr-16 | TCTGTTGTAT CCTCCTGTCG G | GTCTGGAATCG TCGTTAAACG | scaffold-25 | TCG | 5 | 15 | 100 |
| ssr-17 | CTTGGGGTGTC TTITAAAGG G | TGTAGTATAGA GGTAGATAATC GAGCG | scaffold-27 | ATG | 4 | 16 | 152 |
| ssr-192 | TGTCTCAAATAA TGAAGCGGCCG C | GGAACCAACCGCATTTTGC | scaffold-833 | AC | 7 | 14 | 148 |
| ssr-208 | TGACATTATTT CATTGCATAAG GTGGC scaffold-901 | ATTTGAACCTA AGATCACCAGC | AT | 7 | 14 | 118 |
| ssr-27 | TTGACAGAGTC CGGTTTTTTGC | GCGGAATCACCT CTCGACTACG | scaffold-51 | GTT | 7 | 21 | 202 |
| ssr-42 | CTTATCAGAT TGGCTTTGGG C | TTATGGGCCCAT GTAAGCATCC | scaffold-90 | AT | 7 | 14 | 107 |
| ssr-56 | AACATCGAGCGGACCA AAAT CG | TGTATAGCATC CGTCACAATAAG GG | scaffold-141 | AAG | 5 | 15 | 123 |
2.1.1. Analysis of phylogenetic approach

The phylogenetic tree was inferred from a 708-bp fragment of COI from 32 individuals of *E. kamericicus* collected in Indonesia and another related insect COI from Gen Bank NCBI database as reference (Figure 1). There are no previous information about *E. kamericicus* COI sequence in GenBank NCBI database or BOLD database. Therefore, this is the first study using a COI fragment as DNA barcoding technique for Species Identification of this species. The COI sequence result from this study already deposited in the GenBank NCBI database for future study. We used several reference species to build phylogenetic tree this include *Araucarius major* (AY040285), *Amorphocerus talpa* (EU310754), *Abantiadinus nodipennis* (KX191194), *Sphenophorus sp. BYU-CO246* (GU176342), *Naupactus xanthographus* (GU176345), *Hylobitelus xiaoi* (JX847496), *Curculionidae sp. MT-2014* (KM244695), *Curculionidae sp. JH-2016* (KU531398), *Eucryptorrhynchus chinensis* (KP410324), *Eucryptorrhynchus brandti* (KP455482), *Sphenophorus sp. BYU-CO246* (NC_018351), *Rhynchophorus ferrugineus* (NC_028535), *Aegorhinus superciliosus* (NC_027577), *Hylobitelus xiaoi* (NC_022680), *Naupactus xanthographus* (NC_018354), *Apodrosus epipolevatus* (HQ891423), *Laemophloeus fasciatus* (KP134161), *Exophthalmus pictus fulvovirgatus* (KT350641), *Exophthalmus pictus* (KT350642), *Scepticus tigrinus* (LC108870), *Scepticus uniformis* (LC108925), *Meotiorhynchus querendus* (LC108949). Figure 1 shows

![Phylogenetic Tree](image.png)

**Figure 1** Phylogenetic Tree using RAxML method
that the COI fragments from *E.kamerunicus* samples were clustered into one cluster compared with reference from another related species. Based on this result we can conclude that COI gene from *E.kamerunicus* is unique and conserved. This COI gene can be used as DNA barcode for identifying *E.kamerunicus* species, especially using our primers mentioned in this study.

2.1.2. Analysis of network approach

Minimum spanning network (MSN) was used in this study in order to elucidate evolutionary relationships among *E. kamerunicus* cryptic species in Indonesia. The MSN was constructed from 32 haplotypes demonstrated that several haplotypes were highly common and shared by many locations. As we can see in the Figure 2.

The haplotype for one location can be consisted from different locations. From this result we can conclude that there is no genetic bottleneck in sampling population or geographic isolation. The population in this study still in mixed haplotype.

2.2. Nuclear Markers

Understanding population structure and gene flow among regions are very important aspects in the bio-management of this species. In this study, we obtained data from both mitochondrial and nuclear DNA markers of an oil palm pollinator weevil.

The nuclear loci are essential as well, because they represent neutral and paternal inheritance. Particularly, non-coding intron sequences offer potentially powerful genetic markers, since they have a number of traits that are suitable for molecular phylogenies

2.2.1. Analysis of allelic pattern

The result of allelic frequency of each marker can be seen in Table 3. It is showed that the marker used in this study is highly polymorphic.

The number of alleles among all individuals at the 16 selected SSR loci ranged from 3 to 26. The observed heterozygosity (Hobs) ranged between 0.00 to 1.000, expected heterozygosity (Hex) ranged between 0.276 to 0.949 and polymorphism information content (PIC) ranged between 0.248 to 0.930 respectively.

The program CERVUS found significant deviations from HWE for 13 loci and estimated a frequency of null alleles ranging from -0.2707 to 0.1218 for the individual loci. Null alleles can be confirmed through departure from Mendelian segregation ratio in progeny sets. Detection from expected segregation ratios were tested for each locus in each progeny family.

The result of allelic pattern across population can be seen in Figure 3. It is shown that the heterozygosity (He) between population range from 0.35 to 0.65. With the highest was in West Java (Jawa Barat) population and the lowest was in Central Sulawesi (Sulawesi Tengah) population.

![Figure 2 Minimum spanning network (MSN) inferred from samples mitochondrial sequences COI](image)
2.2.2. Clustering analysis

Clustering analysis was conducted in order to find population cluster between sample tested in this study. We used Principal Component Analysis (PCA) and Discriminant analysis of principal components (DAPC). Principal component analysis (PCA) is a statistical procedure that uses an orthogonal transformation to convert a set of observations of possibly correlated variables (entities each of which takes on various

| Locus      | K* | N* | Hobs* | Hexp* | PIC* | HW* | F(Null)* |
|------------|----|----|-------|-------|------|-----|----------|
| Locus-SSR12| 20 | 32 | 1.000 | 0.892 | 0.868 | ND  | -0.0717  |
| Locus-SSR13| 8  | 32 | 1.000 | 0.821 | 0.781 | ND  | -0.1096  |
| Locus-SSR16| 3  | 32 | 0.000 | 0.276 | 0.248 | ND  | 0.9725   |
| Locus-SSR17| 12 | 32 | 1.000 | 0.861 | 0.832 | ND  | -0.0919  |
| Locus-SSR27| 26 | 32 | 0.906 | 0.949 | 0.930 | ND  | 0.0143   |
| Locus-SSR142| 10 | 32 | 1.000 | 0.829 | 0.795 | ND  | -0.1111  |
| Locus-SSR56| 3  | 32 | 0.000 | 0.484 | 0.427 | ND  | 0.9991   |
| Locus-SSR137| 10 | 32 | 0.000 | 0.766 | 0.725 | *** | 1        |
| Locus-SSR159| 12 | 32 | 0.969 | 0.861 | 0.830 | ND  | -0.0748  |
| Locus-SSR192| 12 | 32 | 0.875 | 0.899 | 0.874 | ND  | 0.0023   |
| Locus-SSR208| 6  | 32 | 0.000 | 0.468 | 0.437 | ND  | 0.9987   |

*K = Number of alleles detected; N = Number of individuals, Hobs = Observed Heterozigosity, Hexp = Expected Heterozigosity, HW = Hardy-Weinberg equilibrium test; F(Null) = Null allele frequency estimate

Figure 3 Allelic Pattern across population. He: Heterozigosity, Na: Number of Alleles, Na Freq >= 5% : Number of Alleles more than 5%, Ne : Number of Effective Alleles, I = Shanon Index
numerical values) into a set of values of linearly uncorrelated variables called principal components. The DAPC is a multivariate statistical approach that uses populations defined \textit{a priori} to maximize the variance among populations in the sample by partitioning it into between-population and within-population components. DAPC thus maximizes the discrimination between groups.

As we can see from Figure 4, there are no specific clustering found in sample studied here. From one island to other islands. Based on this result, the population of \textit{E.kamerunicus} was not in geographic isolation. We then look closer to compare one population with another population. It can be seen in Figure 5 that the population genetic from same island was quite similar. For instance, the Sumatran populations genetic were in close distance except Aceh populations.

Minimum spanning networks (MSN) are a great way to visualize relationships among individuals in our data. In can describe the relationship of each individual or population. As we can see from Figure 6 that the individuals from same sites were genetically similar (clustered). Also, we can elaborate the haplotype networking based on this result. It is clear that the haplotype form one site or island was not originated from one haplotype. Therefore, this result supports previous results that the genetic population between islands is not isolated by geographic [44, 45].

\textbf{Figure 4} Principal Component Analysis. Black: Sumatra, Red: Sulawesi, Dark Blue: Kalimantan, Light Blue: Jawa, Green: Papua
2.3. Analysis of Population Structure

In this study we applied two sets of molecular markers to avoid any bias due to the use of only a single marker. In order to find population structure and how many genetic groups in our sample, we conduct population structure analysis. We used the method from [36] and [37] in order to find the best K.

The result from Puechmaille Method showed that the best K was 5 also Evano Method showed that the best K was 7. Therefore, we will use both results to create population structure plots. The result can be seen in Figure 8.

Based on this result, we can see that at least there are 5 or 7 cluster populations from the sample. The only clear cluster populations were from Sulawesi Island. In addition, the genetic from another islands seems admixture from each other. Also based on these results, supporting that we could not find clear geographic isolation between islands and between samples in this study.
Figure 7 The best K estimation using Puechmaille Method (A) and Evanno Method (B).

Figure 8 Population Structure based on K=2 to K=12
~The best K based on Puechmaille Method ; *The best K based on Evanno Method
3. CONCLUSION

This study evidenced that there is high gene flow among the populations of E. kamerunicus in Indonesia based on combined independent molecular markers. This also may have played a role in shaping the observed weak structure detected in this study. In conclusion, E. kamerunicus between islands in Indonesia was not isolated by geographic.

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