

RESEARCH ARTICLE

Genome-wide SNP analyses reveal population structure of *Portunus pelagicus* along Vietnam coastlineBinh Thuy Dang^{1*}, Muhammad Arifur Rahman^{1,2}, Sang Quang Tran¹, Henrik Glenner^{3*}**1** Department of Biology, Institute for Biotechnology and Environment, Nha Trang University, Nha Trang City, Vietnam, **2** Department of Graduate Studies, Nha Trang University, Nha Trang City, Vietnam, **3** Department of Biological Sciences, University of Bergen, Bergen, Norway* binhdt@ntu.edu.vn(BTD); Henrik.Glenner@uib.no (HG)

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Citation: Dang BT, Rahman MA, Tran SQ, Glenner H (2019) Genome-wide SNP analyses reveal population structure of *Portunus pelagicus* along Vietnam coastline. PLoS ONE 14(11): e0224473. <https://doi.org/10.1371/journal.pone.0224473>**Editor:** Arnar Pálsson, University of Iceland, ICELAND**Received:** December 10, 2018**Accepted:** October 15, 2019**Published:** November 5, 2019**Copyright:** © 2019 Dang et al. This is an open access article distributed under the terms of the [Creative Commons Attribution License](https://creativecommons.org/licenses/by/4.0/), which permits unrestricted use, distribution, and reproduction in any medium, provided the original author and source are credited.**Data Availability Statement:** All relevant data are available on the NCBI database. BioProject ID: PRJNA579247 URL: <http://www.ncbi.nlm.nih.gov/bioproject/579247>.**Funding:** This project was funded by NORHED under the framework of research project QZA-0485 SRV-13/0010 titled "Incorporating Climate Change into Ecosystem Approaches to Fisheries and Aquaculture Management in Sri Lanka and Vietnam".**Competing interests:** The authors have declared that no competing interests exist.

Abstract

The blue swimming crab (*Portunus pelagicus* Linnaeus, 1758) is one of the commercially exploited crab fishery resources in Vietnam. This is the first study to provide a broad survey of genetic diversity, population structure and migration patterns of *P. pelagicus* along the Vietnamese coastline. The crab samples were collected from northern, central and southern Vietnam. Here, we used a panel of single nucleotide polymorphisms (SNPs) generated from restriction site-associated DNA sequencing (RADseq). After removing 32 outlier loci, 306 putatively neutral SNPs from 96 individuals were used to assess fine-scale population structure of blue swimming crab. The mean observed heterozygosity (H_o) and expected heterozygosity (H_e) per locus was 0.196 and 0.223, respectively. Pairwise *Fst* and hierarchical AMOVA supported significant differentiation of central and northern from southern populations ($P < 0.01$). Population structure analyses revealed that *P. pelagicus* in the south is a separate fisheries unit from the north and center. Contemporary migration patterns supported high migration between northern and central populations and restricted genetic exchange within the southern population. In contrast, historic gene flow provides strong evidence for single panmictic population. The results are useful for understanding current status of *P. pelagicus* in the wild under an environment changing due to natural and anthropogenic stresses, with implications for fisheries management.

Introduction

The tropical to subtropical Vietnamese coastal zone is divided into the Gulf of Tonkin in the North, the central coast, the southeast coast and the Gulf of Thailand in the South [1,2]. The exclusive economic zone (EEZ) covers about 1 million km² and 3260 km of coastline along the East Sea (the Vietnamese name for the South China Sea). In winter, currents flow in a North East-South West direction while in summer, ocean currents flow from the South West-North East [3,4] with the eddies existing at the southern and central parts of the Vietnamese coastline

[3]. Climate change and human activities including aquaculture, overexploitation, and illegal fishing are threatening coastal habitats (e.g. seagrass beds) and biodiversity [1,5–7].

The blue swimming crab (*Portunus pelagicus*) is a scavenging tropical marine species, widely distributed in the Indian and Pacific oceans, the East coast of Africa, the Mediterranean Sea and southern Japan [7–10]. In Vietnam, it is distributed in the wild throughout the long coastline and aggregated densely in Kien Giang (south of the Mekong Delta) waters [8,11]. It matures and reproduces continuously in one spawning season [12–14]. Planktonic larvae may be transported long distances, supposedly driven by a combination of factors such as temperature, wind, surface currents and salinity [14–17], and spatial distribution depends on larval stages [15,16,18].

P. pelagicus is present in large numbers with great value for commercial fisheries exporting to the USA, Europe and Japan [7,19,20]. According to the FAO (2016) [21], global catch and aquaculture production were 265,896 tonnes, and 29 tonnes, respectively. In Vietnam, the total catch in 2010 was 11,300 tonnes, while production in Kien Giang reached 7,800 tonnes in 2013, suggesting a decline due to overharvesting [11]. Gillnet and crab traps were reported as the dominant fishing gears of *P. pelagicus* (accounting for 77.8% and 22.2%, respectively) [11].

A crab management plan for Vietnam is in place. However, due to unsystematic application of management measures (the minimum landing size and the closed season), and lack of demographic information, management is considered ineffective [22,23]. Recently, genetic studies have increasingly been applied to improve understanding of stock size, gene flow, distribution and migration patterns of subpopulations in mixed fisheries [24–27]. Population information including connectivity across species distribution range, exchange rate and source-sink dynamics are important for understanding potential impacts of bio-physical factors [28–31], human-induced fragmentation [32] and pollution [33,34], or overexploitation [35–37]. Among the wide-range of molecular approaches, restriction site-associated DNA sequencing (RADseq) is well known for its ability to identify and score thousands of single nucleotide polymorphisms (SNPs), which are randomly distributed across the target genomes using next generation sequencing [38–40]. RAD methods are being used and developed with many techniques such as mbRAD [41], 2b-RAD [42], ddRAD [38] and ezRAD [43].

Several studies have revealed different population structuring of *P. pelagicus* throughout its distribution range. In the early 2000s, Yap et al. (2002) [44] and Sezmis et al (2004) [45] detected high population genetic structure in Australia with microsatellites. Similarly, Klinbunga (2007, 2010) [46,47] using DNA polymorphism assays (RADPs and AFLPs) identified strong genetic population structure in Thailand. A more recent studies utilizing mitochondrial DNA markers discovered either limited or high genetic structure in China and the Philippines, respectively [48,49]. In both studies, cryptic species of *P. pelagicus* have been reported as previously recorded by Lai et al (2010) [10]. Additionally, using microsatellites, Chai et al (2016) [50] identified low genetic structure of *P. pelagicus* in Malaysia, while Ren et al. (2016) [51] found distinct populations in Indonesia with RADP plus nuclear DNA marker (16S rDNA). Recently, Miao et al. (2017) [52] applied RADseq to investigate 91 SNPs suggesting these as helpful makers for population research resources of this valuable species. Despite the economic and ecological importance, no studies are known of population genetics of *P. pelagicus* in Vietnam, although limited published studies examining genetic structure of marine organisms have indicated high connectivity in the dynamic and complex Vietnam East Sea waters [29,53].

The goal of this study was to develop SNPs using RADSeq, previously not accomplished for *P. pelagicus* in Vietnam, to better understand fine-scale population structuring and gene flow along the Vietnamese coastline and to provide data on resilience and sustainability for fisheries management.

Materials and methods

Sampling sites and tissue collection

Blue swimming crabs were collected along the north-south geographical temperature gradient: Cat Ba Island—Hai Phong City; Ha Long Bay—Quang Ninh Province (northern population), Nha Trang Bay and Van Phong Bay—Khanh Hoa province; Song Cau and Tuy Hoa—Phu Yen province (central population), and Phu Quoc Island, Rach Gia City—Kien Giang Province (southern population) (Fig 1, Table 1).

The crabs were collected at the exploitation sites, transported alive in aerated sea water to the laboratory where they were kept in aquaria until tissue sampling. Information on sampling sites and crab size (carapace width and weight) were presented in S1 Table. All tissue samples were taken from chelipeds of fresh crab and preserved in 95% ethanol.

Research methodology

DNA extraction and digestion. Genomic DNA was extracted from preserved tissue samples using the DNeasy Blood & Tissue Kit (Qiagen) following the manufacturer's instructions,

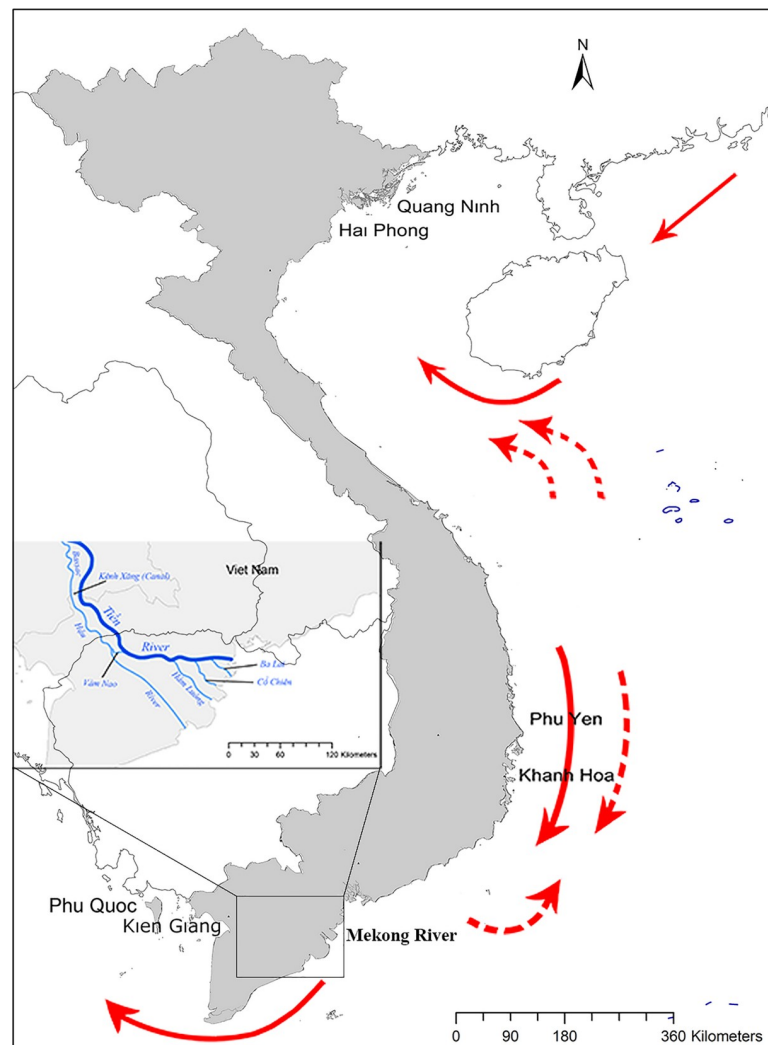


Fig 1. Sampling map of *Portunus pelagicus* and surface currents following northeast (bold line) and southwest (dash line) monsoons, INSET: Mekong River (black box) in Mekong delta, Vietnam.

<https://doi.org/10.1371/journal.pone.0224473.g001>

and treated with RNase (100 mg/mL) to remove residual RNA. Extracted DNA was eluted three times (100 μ l elution/time) to get better DNA quality. All elutions were assessed using gel electrophoresis (1% agarose gel). The best elution (sharp, high weight molecular bands, no smear) was selected to determine the concentration by Qubit[®] 2.0 Fluorometer (Invitrogen). Selected DNA templates (100 ng, concentration \geq 3 ng/ μ l) were then purified using AMPur-XP (Agencourt) beads using a 2:1 template to bead volume ratio with the beads left in.

Purified DNA from each crab individual was simultaneously digested with two restriction enzymes: MboI and Sau3AI (NEB). Each digestion was performed in 25 μ l reactions: 2.5 μ l SmartCut Buffer (10X), 0.5 μ l MboI and 0.5 μ l Sau3AI (5 unit/ μ l), and 21.5 μ l of DNA template (eluate from the beads). Digestions were incubated at 37°C for 3 h to overnight, and then 65°C for 20 min, cleaned with PEG solution (10 g PEG, 7.3 g NaCl, plus water up to 49 ml), and eluted with 20.1 μ l Illumina Resuspension Buffer.

EzRAD library preparation. Cleaned digestions were inserted directly into the Illumina TruSeq nano DNA library Prep kit following the Sample Preparation v2 Guide starting with the “Perform End Repair” step for one-third volume reactions (Supplement S1 [38]). Digested libraries were end-repaired, 350 bp size-selected by SP bead. Firstly, SP bead:H2O (1.5:1) were added to removed $>$ 550 bp fragments, the supernatant collected and applied to 10 μ l SP bead to subsequently remove $<$ 350 bp fragments”. The 3’ ends of selected libraries were then adenylated and Illumina adapters were ligated to the digested genomic DNA samples. PCR reactions were performed using a total volume of 15 μ l including 1.5 μ l Illumina PCR Primer Cocktail, 6 μ l Illumina Enhanced PCR Mix, 1.875 μ l ddH₂O and 5.625 μ l DNA libraries. Biorad thermocyclers (Icycler) were used under the following temperature program: initial denaturation at 95°C for 3 min, followed by 8 cycles of 98°C for 20s, 60°C for 15s and 72°C for 30s. Final extension was done at 72°C for 5 min and the soaking temperature was set to 4°C.

PCR products (The 400–500 bp fragments of which 120 bp are the ligated adapters) were inspected using a 1.5% agarose gel with ethidium bromide and bands were visualized under UV transilluminator. PCR products were purified using SP Beads (1:1), and quantified using qPCR. DNA libraries were sequenced as paired-end 100 bp runs on HiSeq 2500/4000 system (Illumina) in Texas A&M University Corpus Christi Genomics Core Laboratory, USA.

Data analyses

SNPs discovery and filtering. SNP detection was implemented by dDocent v2.0 pipeline [54]. At first, raw FastQ files were trimmed using Trimmomatic v0.3 [55] to simultaneously remove Illumina adapter sequences, and any bases that had a quality score (Q-score) of less than 10 [43]. These reads were clustered and input into *de novo* reference assembly in Rainbow v2.0.2 [56] and CD-HIT v4.6.1 [57,58] based on overall sequence similarity (90% by default). Quality-trimmed reads were mapped to the reference using BWA v0.7.12 [59,60] with the

Table 1. *Portunus pelagicus* sample site information and genetic diversity. Number of individuals successfully genotyped and used in analyses (Nse), observed number of alleles (Na), effective number of alleles (Ne), observed (Ho) and expected (He) heterozygosity, percentage of polymorphic loci (%P) and the inbreeding coefficient (G_{IS}).

Pop ID	Sampling site	Nse	Na	Ne	Ho	He	G _{IS}	%P
Northern	Quang Ninh	16	1.923	1.315	0.166	0.211	0.185	92.31
	Hai Phong	24						
Central	Phu Yen	19	1.982	1.378	0.207	0.246	0.154	98.22
	Khanh Hoa	11						
Southern	Phu Quoc	16	1.885	1.370	0.216	0.233	0.265	88.46
	Rach Gia	10						
	Total/Mean	96	1.930	1.354	0.196	0.230	0.168	93.00

<https://doi.org/10.1371/journal.pone.0224473.t001>

MEM algorithm [61]. SAM files were converted to BAM files using SAMTOOLS [62] and output was further restricted to reads with mapping quality above 10.

SNP calling was performed using Freebayes v0.9.21 [63] with default parameters. Raw SNP files were concatenated into a single variant call format (VCF) file using VCFtools v0.1.11 [64]. The raw SNPs were then filtered with VCFtools and VCFfilter. Primary filtering steps included: minor allele frequency ($MAF > 0.05$), minimum mean depth (≥ 5 mean $DP \leq 10$), INDEL loci (this step decomposed insertion and deletion genotypes), Hardy-Weinberg Equilibrium (HWE with $p < 0.001$), mean quality score ($Q > 30$), max-missing (to apply a genotype call rate of 90% across all individuals), and number of variants (restricted to bi-allelic SNPs). Secondary filtering steps included keeping loci based on allelic balance ($AB > 0.3$), mean mapping quality ($0.9 < MQM/MQMR < 1.05$), and proportion of alternate alleles ($0.05 < PAIRED/PAIREDR < 1.75$). Putative SNPs were submitted to rad_haplotyper (https://github.com/chollenbeck/rad_haplotyper) to remove possible paralogs, and one SNP filtering to get the validated SNP panel.

Outlier loci detection and Linkage-disequilibrium (LD) analysis. Our final filtered panel of SNPs was run in BayeScan v2.1 [65] under default parameter settings to identify loci under divergent or balancing selection. A false discovery rate (FDR) correction of 0.05 was applied [66].

LD was measured as the squared pairwise correlation coefficient between loci (r^2) calculated using the 'LD' function in the R package 'genetics' [67]. Selected outlier clusters (SOC) and Compound outlier clusters (COC) were identified by LD network analysis using R package 'LDna' [68], optimal value of ϕ and |E|min parameter and LD threshold was set up for SOC. LD network were constructed using the R package 'igraph' [69].

All loci putatively identified by either programs were removed from the dataset to generate a panel of neutral SNPs.

Genetic diversity and relatedness. Numbers of alleles (N_a), effective numbers of alleles (N_e), expected (H_e) and observed (H_o) heterozygosity, and inbreeding coefficients (G_{IS}) were calculated for each sampled population and over all populations across the Vietnam coastline using GenAlexv6.5 [70] and GenoDive v.2.0b27 [71].

High levels of relatedness can impact analyses of population structure and estimates of population size, so relationships between individuals were estimated with the R package 'related' [72] using the dyadic [73] and triadic [74] maximum likelihood estimators and allowing for inbreeding. For both estimators 95% confidence intervals were calculated with 500 bootstrap events for each pairwise comparison. Potential pairs were identified as exhibiting a related value. Due to the imbalanced numbers of related pairs among populations leading to reduced sample size and avoiding positive bias in estimates due to underestimating relatedness in the overall population [75], further analyses were run with two datasets, one containing all individuals (with related pairs) and one with one putative individual removed per related pair (related individuals removed).

Analyses of population structure. Pairwise comparisons of F_{st} values between *P. pelagicus* populations were computed in ARLEQUIN [76] to test for significant differentiation among sampled sites. All p-values underwent FDR correction to avoid false positives resulting from multiple comparisons [66]. A hierarchical analysis of molecular variance (AMOVA) was performed to test for significant population structure within species, following two group options: geographically-defined populations (northern, central and southern) and combined individuals from the north and center into a single population, and considering individuals from the south as a separate population (northern-central and southern) using the program ARLEQUIN.

We tested for population connectivity and structure in the program Structure v2.3.4 [77,78] using a model-based Bayesian clustering method to infer the number of lineages, K , in a

dataset. Structure was run to test K values of 1 through 4 with 10,000 iterations of burn-in followed by 5,000 Markov Chain Monte Carlo (MCMC) steps, using the correlated allele frequencies admixture model. The optimal value of K was evaluated using the Evanno method [79] by Structure Harvester v0.6.94 [80]. A Discriminant analysis of principal components (DAPC) was performed using the R package ‘adegenet’ [81]. This analysis provides a graphic description of the genetic divergence among populations in multivariate space.

Migration patterns. Historic gene flow between populations was estimated using the Bayesian inference implemented in MIGRATE-n v3.6.11 [82]. MIGRATE-n’s implementation of coalescent theory measures migration $4 \times Ne$ generations in the past [32,83]. Sample sizes were reduced for each population to obtain 200 loci genotyped in 100% of individuals used for the analysis. The run was performed using 500,000 recorded genealogies sampled every 100 steps, preceded by a burn-in of 20,000. Four hot chains were used with temperatures: $T_1 = 1.0$, $T_2 = 1.5$, $T_3 = 3.0$ and $T_4 = 1.0 \times 10^6$. After optimization, the maximum mutation-scaled effective population size (θ) prior was set at 0.1 while the maximum mutation-scaled migration (M) prior was set at 20,000. Five hypotheses of migration among populations were tested: (1) symmetric migration rates between all sites (Panmixia Model), (2) non-symmetric migration rates between all sites (Full Model) (3) migration between all sites only from the north to the south (North-South Model), (4) migration between all sites only from the south to the north (South-North Model), (5) migration occurring only between neighboring, north-center sites but no migration between south population (South Separate Model). The most likely model was chosen using the Bezier In produced by Migrate-N according to Beerli et al. (2009) [84]. To elucidate the recent migration patterns, estimate relative migration levels (Nm) between populations were calculated based on neutral SNPs using divMigrate function [85] of R package “diveRcity” [86]. Gene flow patterns were visualized using network graphics produced using the R package “qgraph” [87].

Ethics Statement: All crab were collected from fish markets or through normal fishing activities and therefore within the guidelines of approved IACUC procedures, and did not need sampling permission in Vietnam. This study did not involve protected or endangered species

Data Archiving: Upon acceptance, the unmodified sequence data in FASTQ format used in this research along with corresponding metadata will be uploaded and archived in the publicly accessible Genomic Observatories Metadatabase (GeOME, <http://www.geome-db.org/>).

Results

SNP discovery and filtering

Results of 165 libraries of *P. pelagicus* along the Vietnamese coastline generated 604123297 reads with a reading length of 101 bp. The optimal reference assembly of 3280843 bp was constructed from 9583 RAD tags. Initially, 107115 raw SNPs were detected. After filtering steps, 96 individuals were successfully genotyped at 338 valid SNPs. Information on individuals removed and SNPs retained at each step of filtering and data analysis is presented in **S2 Table**.

Outlier loci detection

BayeScan identified thirteen SNPs as outliers ($q < 0.05$, $\alpha > 0$, $FDR \leq 0.05$) from the panel of 338 putative SNPs used to detect selection footprints (**Fig 2D**). LD network presented one selected outlier cluster (SOC) including 32 loci ($\varphi = 1$ and $|E|_{\min} = 30$, $\lambda_{\min} = 0.79$, LD threshold = 0.39) (**Fig 2A–2C**). The outlier loci detected by BayeScan were included in the SOC of LD network. In total, 32 loci were removed from the SNP panel and the 306 remaining loci were assumed to be neutral.

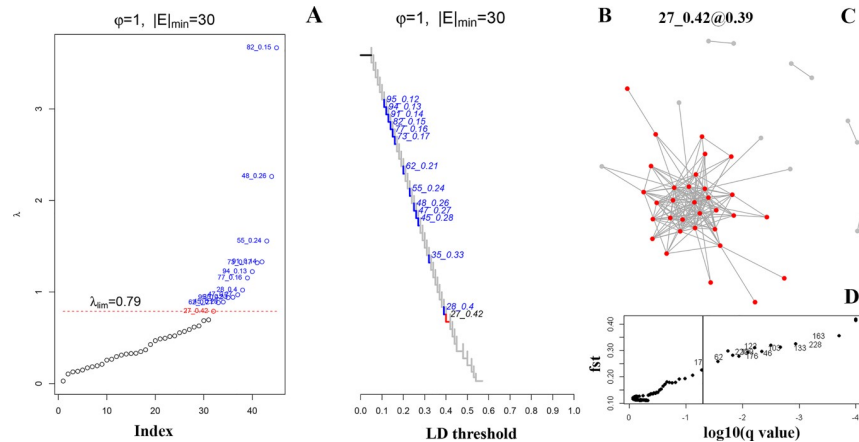


Fig 2. LD network analysis and outlier test results of *Portunus pelagicus*. (A) All λ values in increasing order with values above λ_{\min} corresponding to outlier clusters. Parameter values for ϕ and $|E|_{\min}$ are shown above plots. (B) A clustering tree of pairwise r^2 values from putative 338 SNPs. Branches corresponding to SOCs and COCs are indicated in red and blue, respectively. (C) Selected SOC is shown at an LD threshold where it is joined by a single link to other loci. (D) Results of Bayesian outlier test, locus specific F_{st} coefficient is plotted against $\log_{10}(q \text{ value})$ for the model including selection, the vertical line represents a false discovery threshold of 0.05.

<https://doi.org/10.1371/journal.pone.0224473.g002>

Genetic diversity and relatedness

Genetic diversity of *P. pelagicus* is presented in [Table 1](#). The mean observed number of alleles (N_a) and effective number of alleles (N_e) of the populations were 1.930 and 1.354 respectively. Average observed (H_o) and expected heterozygosity (H_e) were shown across all populations, ranging from 0.166–0.216 (mean 0.196) to 0.211–0.246 (mean 0.23), respectively. Inbreeding coefficients ranged from 0.154 (Center) to 0.265 (South), with an overall G_{IS} for all individuals at 0.168. % Polymorphic sites were highest in the central (98.22%), and lowest in the southern population (88.46%).

Analyses of genetic relationships between individuals revealed 23 pairs of putative half siblings and 5 pairs of putative full sibling ([Table 2](#)) following removal of 16 individuals ([S2 Table](#)). Both full and half siblings (22 pairs) occurred abundantly within the southern population, while the other sibling pairs occurred in remaining sampling sites ([Table 2](#)).

Population structure and migration patterns

AMOVA results ([Table 3](#)) of two hierarchical arrangements (3 populations versus 2 populations) and with two data set (with related pairs and related individuals removed) showed the majority of the variation (80.91–87.5%) in *P. pelagicus* was found within individuals, and highly significant in all cases ($F_{IT} = 0.125\text{--}0.19$, $P < 0.001$). The proportion of variance explained by differences among populations (F_{ST}) were larger in the two-populations (17.43% with related pairs and 15.03% when related individuals removed) than in the three-pops arrangements (11.29% and 8.06%, respectively). It is clear that related individuals contributed to the percentage of variation according to different clustering of populations, however, in all cases the difference were highly significant ($P < 0.001$). With all arrangements and two datasets, among individuals within populations (F_{IS}) differentiation were not significant.

Pairwise F_{st} values between southern population to northern and central populations showed statistically significant genetic differentiation ($P < 0.001$) in all arrangements, and data sets ([Table 4](#)). In three-population clustering, the southern population showed more differentiation with the northern 0.199 with related pairs and 0.181 with related individuals removed)

Table 2. Results of relatedness analysis for two estimators calculated with *related* for pairs of putative siblings. Coefficients of relatedness (*r*) with 95% confidence intervals in parentheses are presented for both the Dyadml likelihood estimator and the trioml likelihood estimator. The most likely relationship for each pair is also shown.

Specimen Pairs	Groupings	Trioml (CI 95%)	Dyadml (CI 95%)	Relationship
KG104/KG119	KGKG	0.303 (0.202–0.413)	0.303 (0.201–0.42)	Half siblings
KG102/KG112	KGKG	0.302 (0.164–0.560)	0.307 (0.201–0.592)	Half siblings
KG104/KG117	KGKG	0.303 (0.207–0.432)	0.309 (0.203–0.429)	Half siblings
KG105/KG118	KGKG	0.310 (0.194–0.380)	0.311 (0.204–0.385)	Half siblings
KG109/KG119	KGKG	0.311(0.206–0.421)	0.311 (0.205–0.420)	Half siblings
KG102/KG110	KGKG	0.306 (0.205–0.407)	0.313 (0.205–0.408)	Half siblings
KG104/KG116	KGKG	0.308 (0.120–0.365)	0.314 (0.206–0.371)	Half siblings
KG103/KG110	KGKG	0.314 (0.203–0.397)	0.322 (0.206–0.406)	Half siblings
QN206/QN213	QNQN	0.323 (0.206–0.414)	0.323 (0.207–0.418)	Half siblings
QN213/HP207	QNHP	0.323 (0.202–0.447)	0.327 (0.209–0.446)	Half siblings
PY111/QN214	PYQN	0.311 (0.214–0.420)	0.332 (0.214–0.420)	Half siblings
KG102/KG109	KGKG	0.325 (0.222–0.435)	0.332 (0.229–0.436)	Half siblings
KG101/KG109	KGKG	0.322 (0.245–0.433)	0.333 (0.230–0.436)	Half siblings
KG103/KG104	KGKG	0.327 (0.232–0.450)	0.334 (0.231–0.449)	Half siblings
KG109/KG117	KGKG	0.313 (0.234–0.483)	0.336 (0.234–0.483)	Half siblings
KG104/KG105	KGKG	0.336 (0.238–0.426)	0.342 (0.237–0.436)	Half siblings
KG103/KG106	KGKG	0.333(0.253–0.430)	0.345 (0.249–0.457)	Half siblings
KG102/KG121	KGKG	0.372 (0.246–0.452)	0.372 (0.262–0.448)	Half siblings
KG110/KG112	KGKG	0.366 (0.269–0.466)	0.374 (0.279–0.465)	Half siblings
KG104/KG111	KGKG	0.375 (0.283–0.482)	0.381 (0.284–0.482)	Half siblings
KH216/QN214	KHQN	0.356 (0.295–0.489)	0.392 (0.306–0.491)	Half siblings
KG103/KG118	KGKG	0.390 (0.316–0.466)	0.402 (0.320–0.477)	Half siblings
KG104/KG118	KGKG	0.444 (0.350–0.540)	0.447 (0.361–0.54)	Half siblings
KG108/KG204	KGKG	0.610 (0.523–0.694)	0.619 (0.535–0.696)	Full siblings
KG122/KG202	KGKG	0.759 (0.696–0.843)	0.762 (0.696–0.842)	Full siblings
KG107/KG201	KGKG	0.783 (0.721–0.851)	0.785 (0.723–0.851)	Full siblings
PY207/PY209	PYPY	0.842 (0.770–0.892)	0.847 (0.769–0.891)	Full siblings
QN219/QN220	QNQN	0.929 (0.856–0.974)	0.941 (0.869–0.973)	Full siblings

<https://doi.org/10.1371/journal.pone.0224473.t002>

than the central (0.143 and 0.117, respectively). However, connectivity was observed between northern and central populations in all cases ($F_{st} = 0.004$, $P = 0.45$ and $F_{st} = 0.0024$, $P = 0.687$).

The STRUCTURE analysis, plotted with a *K* of 2 as chosen by the Evanno method, also showed a clear distinction between the south and the remaining two populations. The similar patterns were observed either with related pairs or related individuals removed from SNPs panels. The southern population was assigned to a first lineage with high certainty (98.4% and 98% composition of the “red” lineage and 1.6% and 2% of “green”. Northern and central populations were assigned to the second lineage with the north represented by a dominance (98.2% and 98%) of “green” lineage, and central exhibiting a mixing of “green/red” with percentages of 82.5/17.5 and 80/20 (Fig 3A left and right).

The Discriminant analysis of principal component (DAPC) showed a clear distinction between the southern population from northern and central populations in both neutral SNPs data sets (Fig 3B). In the dataset with removed related individuals, the northern and central populations were somewhat separated ((Fig 3B, right). However, DAPC analysis based on the 32 under-selected loci showed similar results to neutral related pairs SNPs (Fig 3C).

Table 3. Hierarchical analysis of molecular variance (AMOVA) in *Portunus pelagicus*.

Source of variation	Sum of square	Variant components	% of variation	Fixation index	P value
Three populations (northern, central and southern) with related pairs					
Among populations	81.570	0.580	11.29	$F_{ST} = 0.11$	<0.001
Among individuals within populations	426.065	0.06	1.21	$F_{IS} = 0.01$	0.330
Within individuals	428.000	4.46	87.50	$F_{IT} = 0.125$	<0.001
Three populations (northern, central and southern) with related individuals removed					
Among populations	32.271	0.26	8.06	$F_{ST} = 0.08$	<0.001
Among individuals within populations	237.742	0.14	4.49	$F_{IS} = 0.05$	0.07
Within individuals	224.000	2.80	87.45	$F_{IT} = 0.125$	<0.001
Two populations (northern–central, southern) with related pairs					
Among populations	76.98	0.954	17.43	$F_{ST} = 0.174$	<0.001
Among individuals within populations	430.690	0.662	1.13	$F_{IS} = 0.01$	0.320
Within individuals	428.000	4.46	81.44	$F_{IT} = 0.186$	<0.001
Two populations (northern–central, southern) with related individuals removed					
Among populations	29.708	0.52	15.03	$F_{ST} = 0.15$	<0.001
Among individuals within populations	240.305	0.14	4.06	$F_{IS} = 0.05$	0.07
Within individuals	224.000	2.80	80.91	$F_{IT} = 0.19$	<0.001

<https://doi.org/10.1371/journal.pone.0224473.t003>

Historic migration results strongly supported the Panmixia model based on the highest Bezier approximation score ($\ln = -115475.9$) in which migration was maintained among all sites with random mating between crab individuals (Table 5). The analyses disclosed that there was no mating restriction between crab individuals in the history supposed to be over 1000s of years [32,83]. The populations were able to share genetic material either through larval dispersal due to currents or via migration of adult crabs. Directional migration relative rates among recent *P. pelagicus* populations range from 0.1 to 1 (Fig 3D). Among these, asymmetric directional migration seems to have occurred from southern to northern and central populations, however, bootstrap analysis ($nbs < 0$) showed that directional migration was not significant. Migration from northern to central, however, involved significant asymmetric migration ($nbs > 0$) (S1 Fig).

Discussion

The fine scale population structure of swimming crab, applicable in fisheries management was investigated in both putative neutral and outlier loci. Overall, the current analysis based on SNP panels (including or removing related individuals) all showed similar results. The genetic patterns appear to indicate that *P. pelagicus* in northern-central and southern areas of the

Table 4. Pairwise values of F_{st} (above the diagonal) and their respective P-values (below the diagonal). Bold values indicate significant differences between populations.

Pop ID	With related pairs			With related individuals removed		
	northern	central	southern	northern	central	southern
northern	-	0.0004	0.199	-	0.0024	0.181
central	0.45	-	0.143	0.687	-	0.117
southern	0.000	0.000	-	0.000	0.000	-
	northern-central		southern	northern-central		southern
northern-central	-		0.174	-		0.15
southern	0.000			0.000		

<https://doi.org/10.1371/journal.pone.0224473.t004>

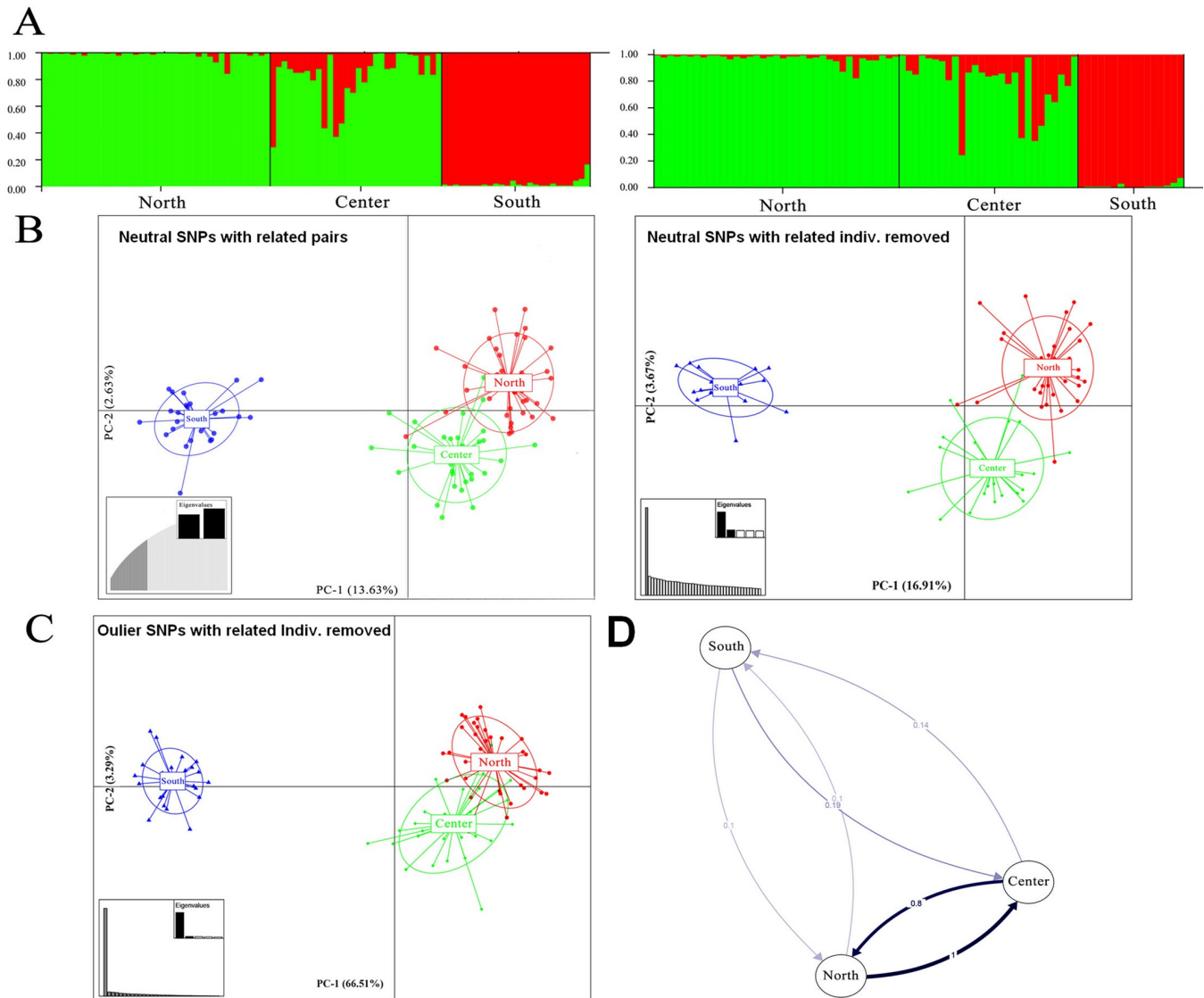


Fig 3. Population structure and migration patterns of *Portunus pelagicus* along the Vietnamese coastline. The bar plot showing individual assignments to inferred clusters (optimal $K = 2$) using the neutral SNP panels (A) with related pairs (left) and related individuals removed (right) in the program STRUCTURE. Each genotype is represented by a single vertical bar. Scatter plot from DAPC following two neutral SNP panels (B) and outlier loci (C), the percentage of variability explained by each coordinate is shown in brackets. The directional relative migration calculated by the divMigrate function performed in the R package diveRsity (D).

<https://doi.org/10.1371/journal.pone.0224473.g003>

Vietnamese coastline maintain distinct populations. Significant pairwise F_{st} comparisons showed strong genetic differentiation between southern to central and northern populations. Furthermore, the hierarchical AMOVA results supported two regional clusters with higher proportions of variation compared to a three-population arrangement (Table 3). Structure and DAPC analyses clearly divided the populations into two subdivisions (northern–central and southern). Outlier SNPs, which represented higher genetic differentiation, and respected providing better resolution to detect fine-scale population structure, identified the same patterns as neutral loci, suggested neutral loci themselves may reflected geographical adaptation ([88]). *P. pelagicus* is well known as a migratory species, both in adult and larval stages. Male

Table 5. Log probabilities of the data given the model (marginal likelihood, based on the Bezier approximation score) and Δ values (difference from largest Lm value) and rank according to largest likelihood value.

Model	BezierLn	Delta	Rank
Full	-117538.33	-2062.47	3
South to North	-129937.02	-14461.16	5
North to South	-116622.97	-1147.11	2
Panmixia	-115475.9	0	1
South separate	-126655.35	-11179.49	4

<https://doi.org/10.1371/journal.pone.0224473.t005>

and female crabs can move between estuaries and open oceans for spawning and/or responding to lowered salinities [7]. As spawning of *P. pelagicus* occurs year-round [11], following the northeast and southwest monsoons, crab larvae may be dispersed by surface currents (Fig 1) along the coast from the Gulf of Tonkin up to the Gulf of Thailand and vice versa. However, all analyses revealed the consistent patterns of non-connectivity from the south to the remaining *P. pelagicus* populations.

The Vietnamese coastline in the East Sea is influenced by seasonally complex water circulations, which result in upwelling and anticyclonic/cyclonic eddies along the south and central coasts [3,4,89]. In general, eddies may limit larval dispersal, acting as a larval retention system [90,91] and maintaining divergence in marine populations [92]. Winds, together with tidal and Mekong river discharge (6000–12000 m³/s) [93,94] were reported as the factors involved in the upwelling, and separate currents in the southern shelf of Vietnam. That may further well explain restricted gene flow in the southern population. Analyses of contemporary gene flow demonstrated the limited genetic exchange in *P. pelagicus* from the south, while the extensive migration occurring along the northern and central coasts. The migration relative rate (Nm) indicated 10 fold greater migration between northern and central populations than from these sites to the south. What makes this more interesting is that significant asymmetric migration from northern to central populations (S1 Fig). Monsoon-induced currents and eddies reported in the central coast [3] make the central region a potential population sink. In contrast, estimates of historical gene flow provided strong evidence for a single panmictic population. This may indicate historical patterns of connectivity were different to those detected today. Vietnamese coasts are currently undergoing dramatic changes due to human activities that heavily affect ecosystems and organisms [1,2]. These human induced disturbances such as overexploitation and habitat degradation/fragmentation as well as coastal pollution may prevent larval transport and dispersal by inducing broad-scale larval mortality [33] and obstructing adult migration [28,95], which may be one of the leading causes of current population isolation.

In term of genetic diversity, the lowest value was detected in the southern population, in concordance with high inbreeding coefficient (0.265) as well as related pairs (Table 2). This heterozygosity deficiency is also recorded in *P. pelagicus* populations in Malaysia [50], and in other marine and freshwater organisms due to widespread habitat loss, degradation and fragmentation [32,96,97]. Significant relatedness and sib-ships have been observed in marine populations due to biophysical larval behavior [98,99], self-recruitment [31,100], and overexploitation/restocking [101]. Kien Giang was the main harvested area of *P. pelagicus* in Vietnam, high level of inbreeding and relatedness, and significant genetic differentiation may indicate that local recruitment originates from a limited pool of successful reproductive adults, and reflect somewhat the pressure of overexploitation on crab populations.

This was the first study to apply the powerful technique of over a hundred SNP markers to infer the natural and/or manmade barriers to gene flow in *Portunus pelagicus*. The population

structure of *P. pelagicus* in the current study does not show high connectivity like other organisms such as lobster [29] and giant clams [53], shown using mitochondrial makers. According to Lemopoulos et al. (2018) [102], RADseq-generate SNPs outperformed microsatellites (and possibly other markers) for investigating individual-level genotypes, and can be applied to studies of small-scale population structure such as the swimming crab in Vietnam. Looking at the swimming crab in the Indo-Pacific region overall, different patterns and levels of genetic structure of *P. pelagicus* have been detected, such as significant genetic differentiation [44,45,47] as well as high gene flow [50,51]. Highly restricted gene flow is mainly reported due to geographic distributions (even at a fine scale) [46], while connectivity is explained by adult migration (such as for spawning), larval dispersal [42], and a lack of physical barriers in the marine environment [50]. In case of Vietnamese *P. pelagicus*, the complex natural and anthropogenic biophysical factors may be driving restricted gene flow along the coastline. However, closely related individuals found in the southern population may affect current results such as reducing the sample size (in the case of related individuals removed), or creating an artificial population structure (when included related pairs). However, the two data sets analyses give the same structure, so we can also confirm an accurate reflection of results for a true phenomenon in this species. *P. pelagicus* can therefore be considered two fisheries and conservation management units. The factors driving current connectivity patterns of *P. pelagicus* are complex, and cannot accurately be identified. *P. pelagicus* is likely at risk from inbreeding and sub-population isolation, and subsequently poor adaptive potential. The management for this species should be careful to ensure that overfishing and habitat degradation do not further affect the vitality of existing populations. Immediate actions such as a seasonal ban on catching crabs in the autumn and late summer to increase successful spawning [32], establishment of marine reserves to reduce genetic losses [101], and coastal pollution control to increase numbers of breeding individuals and larval dispersal [28,33]. Moreover, gear regulation, habitat monitoring and restoration might be one of the most effective ways to manage healthy populations. The appropriate explanation for the high rate of self-recruitment observed in the southern swimming crab remains open. Periodic surveys on genetic diversity, and seascape research [100] should be conducted to provide an overall temporal and spatial view of crab populations.

This study of *P. pelagicus* highlighted the important of conservation genetic studies using advanced genomics for information-lacking geographic zones such as Vietnam East Sea. These results also provide important baseline measures of diversity that can be used for future genetic surveys as well as for monitoring responses of *P. pelagicus* for environmental changes and temperature rises due to climate change.

Supporting information

S1 Fig. Illustrate the significant directional asymmetric migration calculated by divMigrate function performed in R package diveRcity. Number of bootstraps (nbs) were presented along the links.

(TIF)

S1 Table. Sample sites and size of *Portunus pelagicus* with successful sequences, pre-analyzed (*de novo* assembly, mapping) and analyses of population structure. Carapace width (CW) and weight (W). Abbreviation for sampling locations as shown in Table 1.

(DOCX)

S2 Table. Numbers of *P. pelagicus* individuals and SNPs following the filtering steps, outlier, Linkage disequilibrium analysis and relatedness.

(DOCX)

Acknowledgments

We thank the team members of Biodiversity and Conservation, Institute for Biotechnology and Environment, Nha Trang University who helped us collect crab tissues. We are very grateful to Truong Thi Oanh who provided help in data collection. Thanks also to David J. Rees, and Fred Allendorf who helped with linguistic correction of the finished manuscript. Finally, we would like to express our gratitude to three anonymous reviewers for constructive advices and comments that substantially improved the quality of the manuscript.

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