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Using a butterflyfish genome as a general tool for RAD-Seq studies in specialized reef fish

Running title: **RAD-Seq in butterflyfish**

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Abstract

Data from a large-scale restriction site associated DNA (RAD-Seq) study of nine butterflyfish species in the Red Sea and Arabian Sea provided a means to test the utility of a recently published draft genome (*Chaetodon austriacus*) and assess apparent bias in this method of isolating nuclear loci. We here processed double-digest restriction-site (ddRAD) associated DNA sequencing data to identify single nucleotide polymorphism (SNP) markers and their associated function with and without our reference genome to see if it improves the quality of RAD-Seq markers. Our analyses indicate (1) a modest gap between the number of non-annotated versus annotated SNPs across all species, (2) an advantage of using genomic resources for closely related but not distantly related butterflyfish species based on the ability to assign putative gene function to SNPs, and (3) an enrichment of genes among sister butterflyfish taxa related to calcium transmembrane transport and binding. The latter result highlights the potential for this approach to reveal insights into adaptive mechanisms in populations inhabiting challenging coral reef environments such as the Red Sea, Arabian Sea, and Arabian Gulf with further study.

Introduction

Ecologists and evolutionary biologists often mandate genetic data to test hypotheses related to recent population history (i.e. bottlenecks), connectivity among sites, and the spatial distribution of genetic variation for effective species management. Moreover, the capacity to rapidly genotype many individuals' at large numbers of genetic markers improves our ability to estimate demographic parameters (e.g. gene flow, effective population size, admixture), resolve phylogenetic placement, identify genes that may be under selection, and even identify

genes that facilitate adaptation to our rapidly changing environment. Now, with the decreasing cost of high-throughput next-generation sequencing (NGS), data can be sampled from across the genome to meet this demand (for review see Kosuri & Church 2014).

Restriction site associated DNA sequencing (RAD-Seq) methods, which use NGS to target sequence data adjacent to restriction enzyme recognition sites (Davey *et al.* 2011), have generated informative single nucleotide polymorphism (SNP) datasets in several fish species. Examples include studies characterizing different trout species and their hybrids (Hand *et al.* 2015), the genetic basis for different forms of stickleback (e.g. oceanic versus freshwater; Hohenlohe *et al.* 2010; Catchen *et al.* 2013a), identifying cryptic lineages of herring in the Baltic Sea (*Clupea harengus*; Corander *et al.* 2013), and resolving relationships among iconic African cichlid species (Wagner *et al.* 2013; Henning *et al.* 2014). Most of these taxa, however, are considered “model organisms”, and therefore unique in their ability to map, align, and otherwise trace their SNPs back to annotated genomes.

Reef fish, on the other hand, have received much less attention, despite 5000 species inhabiting tropical seas. We note that RAD-Seq methods have only been applied a few times in reef fish, including surgeonfish (Gaither *et al.* 2015), clownfish (Saenz-Agudelo *et al.* 2015), hamlets (Puebla *et al.* 2014; Picq *et al.* 2016), angelfish (Tariel *et al.* 2016), grunts (Bernal *et al.* 2016), groupers (Jackson *et al.* 2014), and parrotfish (Stockwell *et al.* 2016). Most of these studies, however, are limited in scope (based on small sample sizes, few study species, and minimal sampling sites) or fail to confidently assign gene function to SNPs of interest; all rely exclusively on *de novo* assembly of raw sequence reads. This deficiency in reef fish relative to other aquatic organisms can therefore be, at least in part, attributed to a lack of publicly available genomic resources. All published genomes to date in this group are based on cold water (*Takifugu rubripes*; van de Peer 2004), brackish water (*Tetraodon*

nigroviridis; van de Peer 2004), pelagic (*Thunnus orientalis*; Nakamura *et al.* 2013), and phylogenetically distinct (*Rhincodon typus*; Read *et al.* 2015) species.

A recently published genome for the Red Sea butterflyfish (*Chaetodon austriacus*; DiBattista *et al.* 2016a), an obligate corallivore often targeted by the ornamental fish trade (Wabnitz 2003; Lawton *et al.* 2013), allows us to explore genes underpinning ecological niche selection (Cole *et al.* 2008), evolutionary distinctness (Bellwood *et al.* 2010), but more importantly biogeographic patterns related to the specialized and speciose Chaetodontidae family (> 130 species). Indeed, the Red Sea has experienced intermittent historical isolation (Bailey 2009), fluctuating environmental conditions (e.g. Raitso *et al.* 2013), and harbors large numbers of endemic species (DiBattista *et al.* 2016b) whose origins are still subject to debate (DiBattista *et al.* 2016c). The *C. austriacus* genome may therefore enhance RAD-seq analysis in related species.

We therefore have two methodological aims in this study. First, we test whether using the *C. austriacus* genome as a reference improves the quality of RAD-Seq markers for this species, which includes assessing the apparent bias in our method of isolating nuclear loci. Second, to increase the utility of this genome for broader topics of study, we further test the capacity of this genome to serve as a scaffold for RAD-Seq markers in other butterflyfish species. This is the first time comparisons between RAD-Seq analyses, with and without a representative genome, have been conducted across such a large number of related, aquatic organisms.

Material and methods

Sample collection

We collected fin or gill tissue from between 48 and 108 individuals for each of nine species of butterflyfish sampled at sites in the Red Sea and Arabian Sea, including the genomically enabled *C. austriacus* (see Fig.1 and Table 1). We also collected samples from a surgeonfish

species (*Ctenochaetus striatus*; $N = 120$ from 10 populations), which served as an outgroup for our analyses. All samples were preserved in 95% ethanol prior to processing.

Restriction site associated DNA sequencing method

The RAD-Seq method involves digesting genomic DNA with restriction enzymes and sequencing fragments of DNA adjacent to restriction sites. In our case, we used the double-digest RAD tag method (ddRAD; Peterson *et al.* 2012), which uses one common and one rare restriction enzyme cutter but no shearing step, although several other methods exist (Willette *et al.* 2014; Andrews *et al.* 2016).

In brief, genomic DNA was extracted using a Qiagen DNeasy Blood and Tissue Kit (Qiagen, Valencia, CA) following the manufacturer's protocol. Total DNA from each extracted aliquot was quantified using a Qubit dsDNA HS Assay Kit (Invitrogen, Carlsbad, CA). Genomic libraries were prepared from 500 ng of DNA per sample by: 1) digesting at 37°C using the restriction enzymes *SphI* and *MluCI* (NEB), 2) ligating to unique combinations of custom adaptors, 3) pooling 16 individuals at a time, 4) size-selecting 300 to 500 bp fragments on agarose gels from each pool, 5) amplifying over 10 PCR cycles to reduce clonality in 50 μ l reactions containing 25 μ l Illumina True Seq Master Mix, 20 μ l of library DNA, and unique indexing primers that correspond to the standard Illumina multiplexed sequencing protocol, and 6) combining pools, as appropriate, in equimolar concentration to form nine genomic libraries, which were then run on nine lanes of an Illumina HiSeq2000 (101 bp; v3 reagents) at the King Abdullah University of Science and Technology (KAUST) Bioscience Core Laboratory using the SE option.

RAD-Seq analysis

RAD sequences were first analyzed *de novo* (without a reference genome) using Stacks *vers.* 1.44 (Catchen *et al.* 2011, 2013b). Raw reads for each individual in the RAD tag library were trimmed at the end to a common length of 81 bp in FASTQ format using the “process_radtags” pipeline since read quality was considerably lower from this position onwards; reads with an average phred score < 20 (in a 5 bp sliding window) were discarded. All remaining reads from each species were separately analyzed and SNPs were genotyped using the “denovo_map.pl” pipeline. The parameters used in Stacks were: minimum number of reads required to form a stack (m) set to 3; maximum number of mismatches between loci for individuals (M) set to 4; maximum number of mismatches when aligning secondary reads to primary stacks (N) set to 2; maximum number of mismatches between loci when creating a catalog (n) set to 2. This parameter combination yielded the closest number of SNPs for our focal species (*C. austriacus*) compared to the number of SNPs recovered using the “ref_map.pl” pipeline in Stacks (see below and Appendix S1), and was therefore used as the optimal settings for all the other species. Population filtering parameters used in order to include a locus in the final data set were: 1) minor allele frequency > 0.05 and 2) the locus had to be genotyped in at least 80% of individuals (populations: -r) and present in all (strict quality filter) or all but one (relaxed quality filter) populations (populations: -p)(for more detail see Saenz-Agudelo *et al.* 2015).

RAD sequences were also analyzed using the *C. austriacus* genome as a reference (DiBattista *et al.* 2016a). For this approach, trimmed RAD reads from each species (and thus SNPs) were first aligned to the *C. austriacus* scaffolds using Bowtie *vers.* 1.0.1 (Langmead *et al.* 2009). We configured Bowtie to report only one (i.e. the best) alignment per read and allowed up to 3 mismatches in the default seed length. Aligned reads were then analyzed with `ref_map.pl` using the same minimum number of reads to build a stack as outlined above for `denovo_map.pl`; population filtering parameters also remained the same (see Appendix S1). This step additionally allowed us to estimate homology levels for all studied species against the *C. austriacus* genome.

Complimentary Maximum Likelihood (ML) phylogenetic trees were built with RAD sequences using an unpartitioned GTRCAT approximation model in RAxML *vers.* 2.0 (Stamatakis 2006) as implemented in Geneious *vers.* 10.0.2 (Drummond *et al.* 2009)(also see Ree & Hipp 2015). The input for these analyses consisted of a full sequence (.phylip) alignment generated in STACKS (`denovo_map: -m 3 -M 4 -N 2 -n 6`) by compiling four individuals per species, resulting in 38,070 total bp representing 2846 shared SNP loci (2170 fixed SNPs and 676 variable SNPs within species). Only loci present in all 10 species and at least 3 individuals per species were included. Branch support was based on 100,000 rapid-bootstrap replicates with *Ct. striatus* used as an outgroup. Trees were visualized using FigTree *vers.* 1.4.2 (<http://tree.bio.ed.ac.uk/software/figtree/>).

We next explored the form and function of SNPs that mapped to the *C. austriacus* scaffolds in several ways. First, we identified SNPs from *C. austriacus* and its closest relative, *C. melapterus*, which mapped to regions containing gene models. These represent the only two species where using the *C. austriacus* reference genome increased the number of SNPs identified (see below). The associated genes were then used to look for potential enrichment of specific gene functions using GO enrichment analyses. We identified 3775 SNPs from the *C. austriacus* dataset mapping to 3056 different genes in the reference genome of *C. austriacus*, and 2341 SNPs from the *C. melapterus* dataset mapping to 2011 different genes of the reference genome. GO enrichment analyses were performed using TopGo *vers.* 2.22.0 from the R Bioconductor package (Alexa & Rahnenfuhrer 2010), the weight01 model, a cut-off of $P < 0.01$, and specific background gene sets covered by stacks from the `ref_map.pl` analysis. These analyses were used to find which GO terms were over-represented (or under-represented) using annotations for that gene set. Second, we estimated the number of SNPs from *C. austriacus*, with corresponding tolerance intervals (95% confidence in 99% of the sample), that mapped to the assembled genome as a function of scaffold size. This particular analysis highlights scaffolds that host a statistically higher number of SNPs for their length. Frequency distributions for the number of SNPs and scaffold lengths were also calculated.

Results

We used Stacks to assemble and genotype SNPs in two different ways. First, we used `denovo_map.pl` to assemble and genotype SNPs *de novo* for ten different species, including *C. austriacus*, the only butterflyfish with a fully sequenced genome. We also used the `ref_map.pl` pipeline to assemble and map reads from each of these species to the *C. austriacus* scaffolds, which are now publicly available (NCBI Bioproject PRJNA292048; <http://caus.reefgenomics.org>; see Liew *et al.* 2016). Using an optimal parameter combination (see Appendix S1) and `denovo_map.pl`, we found 289,504 loci, with

8,842 (strict quality filter) or 10,711 (relaxed quality filter) variable loci containing at least one SNP site within or between individuals for *C. austriacus* (Table 1). On average, there were 262,247 loci across all butterflyfish species (range: 189,554 to 363,221), and between 476 (*Chaetodon trifascialis*) and 10,257 (*Chaetodon paucifasciatus*) loci per species passing our filters based on the strict criteria, and between 1,271 (*Chaetodon trifascialis*) and 13,539 (*Chaetodon paucifasciatus*) loci per species passing our filters based on the relaxed criteria (Table 1). Average depth of coverage ranged from 10.2X (*Chaetodon fasciatus*) to 16.2X (*Chaetodon paucifasciatus*) for *denovo_map.pl* (average depth of coverage = 12.3X), and from 12.1X (*Chaetodon fasciatus*) to 16.1X (*Chaetodon paucifasciatus*) for *ref_map.pl* (average depth of coverage = 13.5X).

We note a 38% (strict quality filter) or 56% (relaxed quality filter) increase in the number SNPs identified using *ref_map.pl* versus *denovo_map.pl* for *C. melapterus* (Table 1), the closest relative to *C. austriacus*. In contrast, there was a 2- to 6-fold decrease (strict quality filter), or 4- to 12-fold decrease (relaxed quality filter) in the number of SNPs identified for the remaining butterflyfish species using *ref_map.pl* (excluding *C. austriacus*; Table 1). This downward trend was more pronounced when we considered fish outside of the Chaetodontidae family (i.e. *Ct. striatus*, family Acanthuridae), with a 52-fold (strict quality filter) or a 56-fold (relaxed quality filter) decrease in the number of SNPs identified. Indeed, our surgeonfish outgroup, *Ct. striatus*, had only 8 (strict quality filter) or 27 (relaxed quality filter) useable SNP loci after filtering when using *ref_map.pl*.

Gene function under putative selection

In order to determine if variable SNPs in *C. austriacus* and its closest relative, *C. melapterus*, were associated with specific biological processes, we performed GO enrichment analyses on SNPs lying within gene models. To this end, we first analysed if SNPs were randomly distributed with respect to gene function. Analysis of the GO terms covered by the species-specific SNPs showed a strong bias towards particular gene functions, with the strongest enrichment for genes involved in the positive regulation of GTPase activity, axon guidance, and chloride transmembrane transport, among others.

To determine if this observed bias in gene function was a consequence of our RAD-seq method and thus chosen restriction enzyme, we analysed the gene models covered by our loci in comparison to all proteins. Briefly, the *SphI* recognition site (GCATGC) contains an ATG followed by a C, which in coding sequences translates to a methionine (M) followed by a leucine (L). Consequently, using this restriction enzyme could enrich for genomic regions encoding proteins starting with or harboring internal Methionine-Leucine (ML) amino acid stretches. We therefore analyzed the frequency of ML amino acid stretches in all gene models for *C. austriacus* and compared them to the frequency within the set of gene models associated with our loci. This analysis revealed a highly significant enrichment of such genes within our loci, thus identifying this as a source of the previously observed bias (Chi-square test: $P < 0.0001$). Based on this result we generated species-specific GO term backgrounds for our subsequent enrichment analyses to account for the observed bias, which may be an important consideration for other RAD-seq approaches.

SNP specific GO enrichment analyses identified 29 and 37 biological processes in *C. austriacus* and *C. melapterus*, respectively, that were significantly enriched ($P < 0.01$), of which six were shared between the species. These shared terms included calcium ion transmembrane transport, axon

guidance, neuron cell-cell adhesion, positive regulation of excitatory postsynaptic potential, and positive regulation of cardiac muscle cell proliferation. Variable SNPs exclusively found in *C. austriacus* were predominantly associated with genes involved in response to regulation of cytosolic calcium concentrations, morphogenesis, and locomotory behavior (Figure 2a), whereas SNPs exclusively found in *C. melapterus* were enriched for functions involved in ion transmembrane transport, synaptic transmission, and social behavior (Figure 2b). Moreover, we found that only 385 SNPs (i.e. 4%) identified with the *denovo_map.pl* for *C. austriacus* approach failed to map to the reference genome, indicating that both approaches produced representative SNPs.

We next mapped SNP loci identified from *C. austriacus* with *ref_map.pl* to the gap-closed scaffolds in order to track the distribution of SNP density across its own genome. We found that (as expected) the number of SNPs mapping to the reference genome increased with scaffold size ($R^2 = 0.71961$; Fig. 3c). That said, 3055 scaffolds had < 10 SNPs that mapped back to them versus 115 scaffolds with > 10 SNPs that mapped back to them (Fig. 3a) despite a large mean scaffold size (~ 150 kb; Fig. 3b). These two categories (i.e. > 10 SNPs per scaffold versus < 10 SNPs per scaffold) represent sets of SNPs above and below the tolerance threshold, respectively, with a clear difference in GO enrichment between them (see Appendix S2).

Phylogenetic overview

In order to test the phylogenetic generality of these analyses, we assigned SNPs identified within each butterflyfish species, and the outgroup surgeonfish, to annotated versus non-annotated regions of the *C. austriacus* scaffolds. We found that approximately 16% more SNPs were assigned to non-annotated versus annotated regions of the genome, and that the absolute number of SNPs assigned varied among species (Fig. 4a). An estimate of the percent homology to the *C. austriacus* reference genome for all RAD data shows that *C. austriacus* and its closest relative, *C. melapterus*, mapped at a level of 86% and 87% respectively, with the rest of the butterflyfish mapping between 19% and 28% similarity; the outgroup surgeonfish had the lowest level of homology (~2%) (Fig. 4b). This phylogenetic decay is consistent with the inferred relationships among butterflyfish species revealed by a Maximum Likelihood consensus tree built with shared SNP loci (Fig. 4c).

Discussion

We compared analysis of RAD-Seq data using a *de novo* approach (i.e. without a genome) to analysis using a reference genome for a series of Red Sea and Arabian Sea butterflyfish. Our analyses indicate (1) a modest gap between the number of non-annotated versus annotated SNPs across all species, (2) an advantage of using genomic resources for closely related but not distantly related butterflyfish species based on the ability to assign putative gene function to SNPs, and (3) an enrichment of genes among sister butterflyfish taxa related to calcium

transmembrane transport and binding. The RAD-Seq dataset we present here thus provides the most comprehensive estimate of genetic polymorphism to date for any reef fish, and further creates the potential to elevate the Chaetodontidae family to a ‘model group’ for future genomic studies. Overall, our results suggest that RAD-Seq approaches inherently provide valuable insight to population genomic questions for non-model organisms (i.e. those lacking reference genomes), but that detailed analysis of RAD-Seq data is improved when a reference genome is available from a closely related species (as per Pecoraro *et al.* 2015) or when apparent bias in the method of SNP isolation can be identified and accounted for.

Most RAD tags are thought to be randomly distributed across the genome based on the stochastic placement of restriction cut sites (Davey *et al.* 2013), and indeed we found no evidence for what we here refer to as SNP “hotspots” (e.g. Myers *et al.* 2005; also see Fig. 3). The genome assembly of *C. austriacus*, however, consists of gap-closed scaffolds and does not include a linkage map (DiBattista *et al.* 2016a); the exact location of each SNP relative to autosomal (or sex-linked) chromosomes is therefore not yet known. Moreover, we enabled the “select one SNP per read” filtering option in Stacks in order to comply with population genetic assumptions of independent loci, our intended downstream application for these data, which further reduced our ability to detect multiple SNPs sitting within 81 bp of each other. Despite this uncertainty, a large portion of the identified SNPs could be assigned to annotated regions of the genome for most of the butterflyfish species we considered (average number of annotated SNPs = 1199; Fig. 4a). We did, however, detect a strong bias with respect to covered gene functions (i.e. proteins containing ML stretches), which apparently stems from our choice of restriction enzyme. Restriction enzymes that target gene rich regions (e.g. Roda *et al.* 2013), or even specific primers flanking variable SNPs (Campbell *et al.* 2014), have previously been used when more specificity is required. This refinement provides an advantage for researchers who may wish to select a smaller number of SNP sites for genotyping with known spacing, location or function, but also a direct application to projects estimating connectivity and adaptability of fishes. However, such approaches are likely to introduce their own biases that need to be considered in subsequent analyses as we have shown here. Moreover, if a mutation has occurred in the cut-site of some (but not all) individuals, allelic dropout can occur, which has been shown to further bias the inference of genetic variability (Gautier *et al.* 2013). Without an available reference genome, these types of methodological biases are often missed.

We observed a clear advantage to using genomic resources for closely related but not distantly related butterflyfish species in terms of the ability to assign SNPs a putative function. This advantage would likely extend to other sister taxa of reef fish, and as such it would be ideal for studies focused on between-species comparisons *within* monophyletic lineages. Similar advantages were recently shown for the Pacific bluefin tuna (*Thunnus orientalis*; Pecoraro *et al.* 2015). For example, in our case, the availability of the *C. austriacus* genome sequence allowed for functional interrogation of the data using GO annotations within sister taxa, which revealed that variable SNPs were

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significantly enriched for certain gene functions (Fig. 2). These functions included genes associated with neurological processes, which might indicate potential differences in behavior or sensory perception for these specialized reef fish *within* or *between* the different species, and therefore represent plausible candidate genes for future functional interrogation. This is consistent with the hypothesis that detecting and interpreting visual and olfactory cues are highly important for butterflyfish given that most of these fish are monogamous and need to maintain pair bonds, recognize mates, and defend territories from conspecifics (e.g. Boyle & Tricas 2014). More importantly we found that genes associated with calcium transmembrane transport and binding were significantly overrepresented in both *C. austriacus* and *C. melapterus*, suggesting potential convergent adaptations to the prevailing Red Sea and Arabian Sea environments, respectively. A plausible alternative is that these SNPs are not themselves adaptive but linked to other regions of the genome that were not sequenced, resulting in a consistent association and statistically significant enrichment of sequence variation and specific gene functions. If we assume that these functions are under selection, calcium/chlorine ratios are remarkably similar in the Red Sea and adjacent Indian Ocean (Krumgalz 1982), although the Gulf of Aqaba is much lower, suggesting that adaptation to these conditions may dictate which gene functions are conserved. The assimilation of calcium from seawater is an important process for the formation of structural skeleton in corals, protective shells for many of the marine invertebrates, as well as promotes optimal larval development in fish via ossification and gill formation (e.g. Malvezzi *et al.* 2015). Further study of additional fish from the Red Sea and Arabian Sea region could test the hypothesis that genetic adaptations allow species to handle the unique chemical challenges present in regional waters, and particularly could provide some insight to the mechanisms underlying high rates of endemism in the region.

We note that the percentage of reads mapping to the reference genome was actually higher for *C. melapterus* (87.5%) versus the genomically enabled *C. austriacus* (86.4%)(Fig. 4b). Although this result may appear unusual, these species are closely related to each other (Fig. 4c), have only recently diverged (~50 Kya), share mitochondrial DNA haplotypes (see Waldrop *et al.* 2016), and likely hybridize in regions of overlap based on similar behavior observed in nearby seas (DiBattista *et al.* 2015). Recent and/or on-going genetic exchange suggests that introgression across the nuclear genome must be high for these two species (see Waldrop *et al.* 2016). Moreover, despite phylogenetic similarity, *C. melapterus* had lower levels of polymorphism (0.78% versus 2.3%, respectively) and fewer restriction sites recovered (5913 versus 8812 variable loci passing strict criteria, respectively) compared to *C. austriacus*, and yet the average depth of coverage for the former increased from 12.2X to 14.9X when mapping to the reference genome (Table 1). We may therefore have achieved marginally better mapping results for *C. melapterus* simply because we had a lower probability of discarding reads with `ref_map.pl` based on increased homogeneity and higher per locus coverage. We additionally attribute the large range of pre- and post-filtered loci among species to differences in sample size, in addition to stochastic factors such as variable DNA quality for each sample and library preparation protocol.

Conclusion

The rapid identification of a large number of SNPs within populations likewise holds promise for a number of evolutionary-oriented studies. While the number of publicly available genomes is steadily increasing, our results suggest that increasing phylogenetic distance decreases the utility of genomes, but that this remains an important means to identify apparent bias in SNP isolation approaches as well as assess conserved biological processes. Substantial advances in our understanding of evolutionary mechanisms may therefore be possible without the substantial investment of resources required for full comparative genomic studies.

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Data Accessibility

Raw RAD reads (FASTQ format) are available in the NCBI repository, BioProject:

PRJNA292048.

Genome browser URL: <http://caus.reefgenomics.org>

Filtered RAD reads in .vcf format are available from Dryad: doi:10.5061/dryad.f09rh.

Author contributions

J.D.D., P.S.-S., and M.L.B. conceived of and designed the RAD-Seq study. J.D.D., P.S.-S., and M.J.P. produced and analyzed the RAD libraries. X.W. and M.A. performed GO enrichment analyses. All authors developed the manuscript and approve of the final paper.

Fig. 1 Map of the Red Sea and Arabian Sea, including study species and collection sites.

Fig. 2 Top ten highest enriched ($P < 0.01$) gene ontologies for (a) *Chaetodon austriacus* and (b) *C. melapterus* based on annotated SNPs produced using a ddRAD protocol that have protein and GO information available. Gene ontology categories include molecular function, biological process, and cellular component.

Fig. 3 (a) Frequency of SNPs per scaffold for *Chaetodon austriacus* mapping to the assembled genome, (b) frequency distribution of scaffold size (in million base pairs), and (c) number of SNPs as a function of scaffold size using a ddRAD protocol and “ref_map.pl” option in Stacks. Mean = black dashed line; + 2 standard deviations = green dashed lines; tolerance interval = red dashed lines.

Fig. 4 (a) Number of annotated versus non-annotated SNPs (total number of SNPs, i.e. annotated plus non-annotated, is in brackets), (b) percentage of reads mapping to the *Chaetodon austriacus* reference genome using Bowtie for Red Sea and Arabian Sea resident butterflyfish (and an outgroup surgeonfish), and (c) Maximum Likelihood (ML) phylogenetic relationship among species. Mapping parameters used for Bowtie were $-n = 3$ and $-k = 1$. A SNP was considered annotated if located in any region of the scaffold identified as protein coding; a non-annotated SNP was any SNP that did not meet this criteria.

Fig. 1

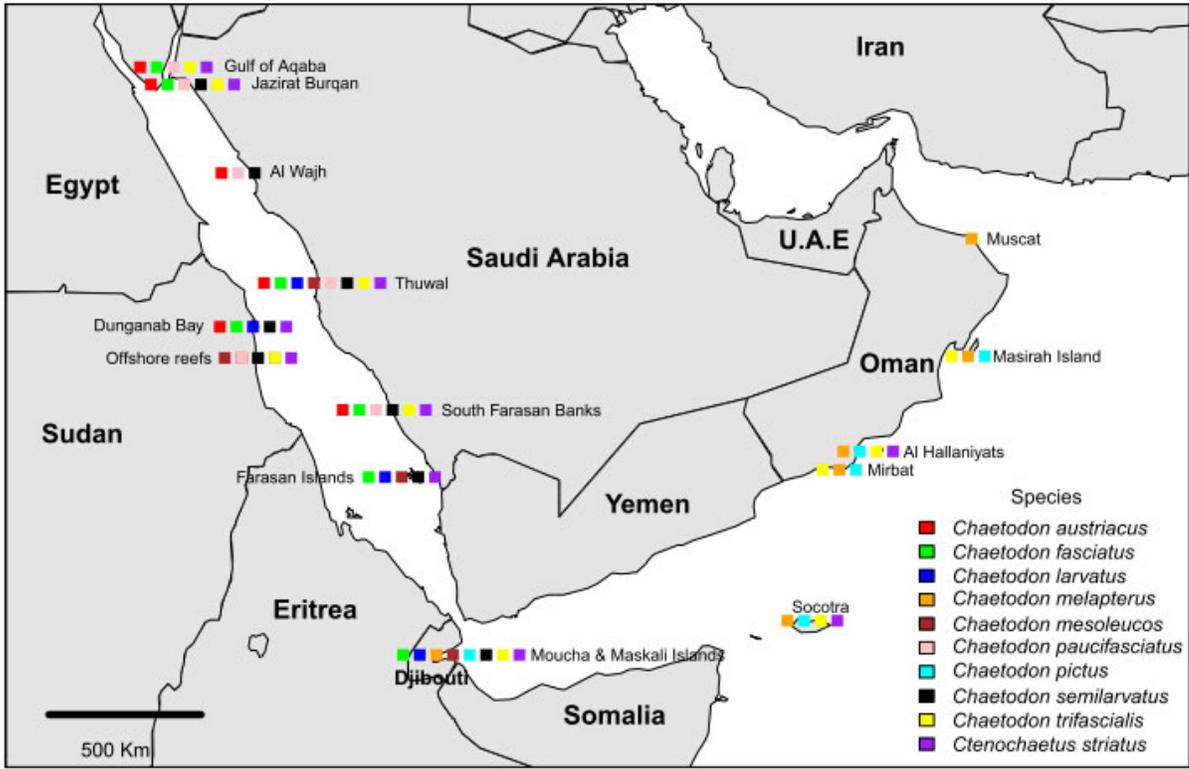


Fig. 2a

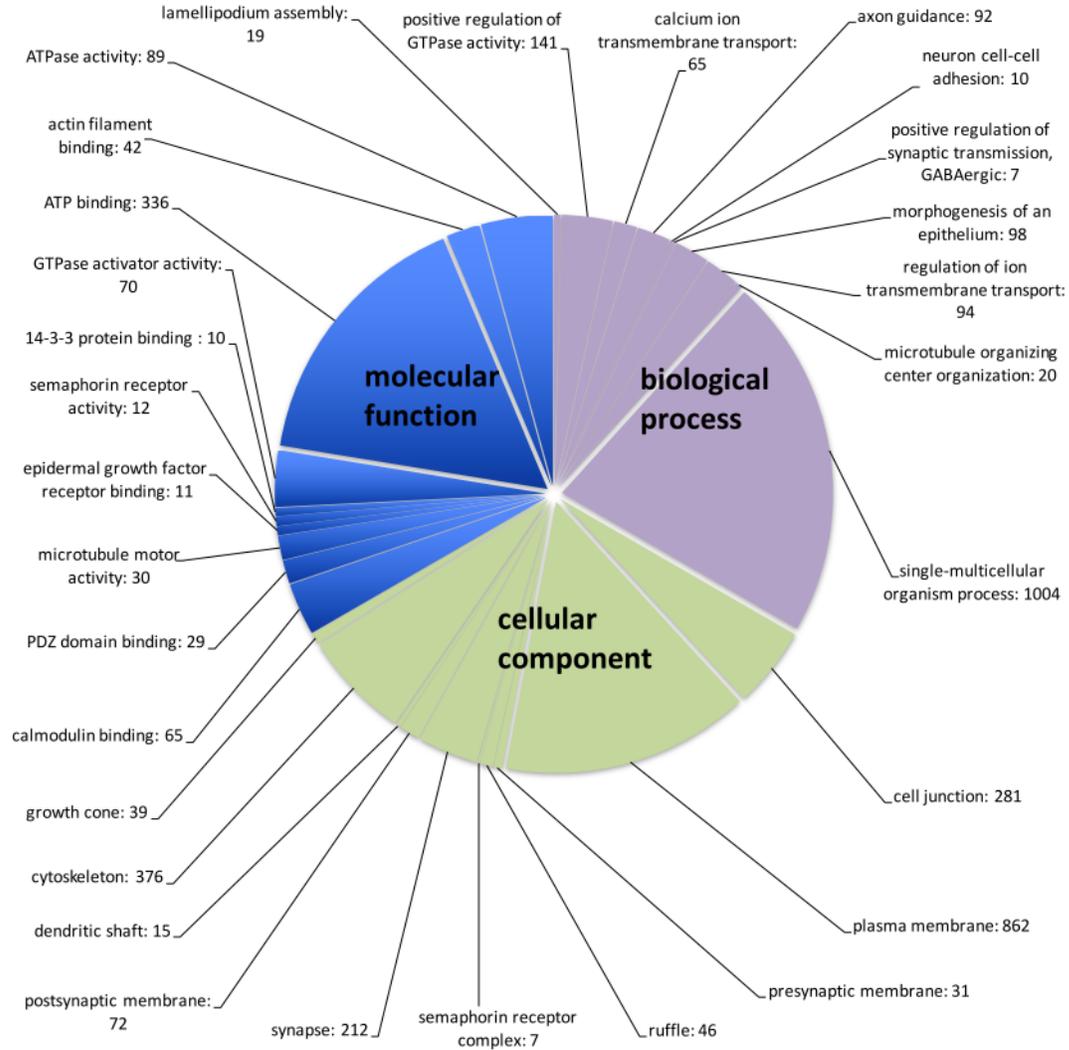


Fig. 2b

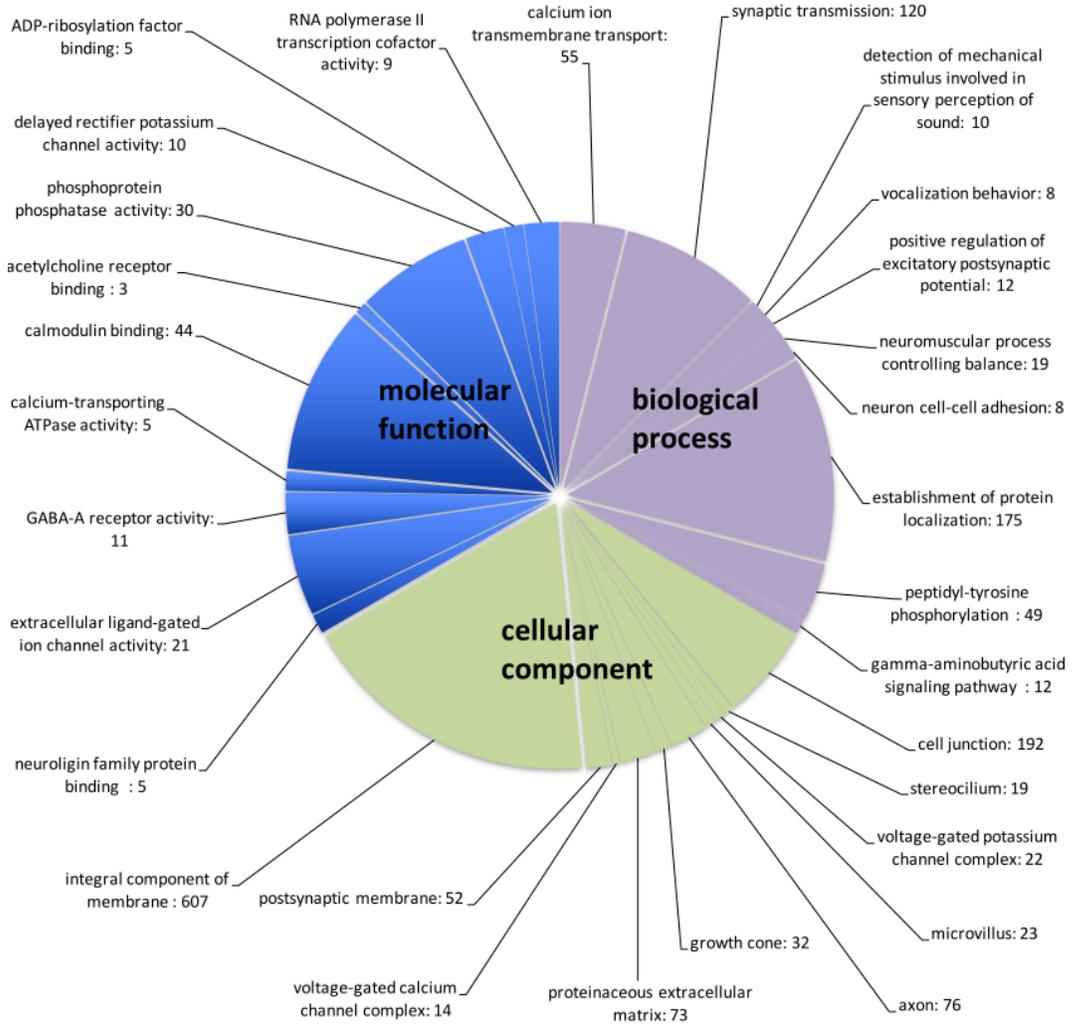


Fig. 3a

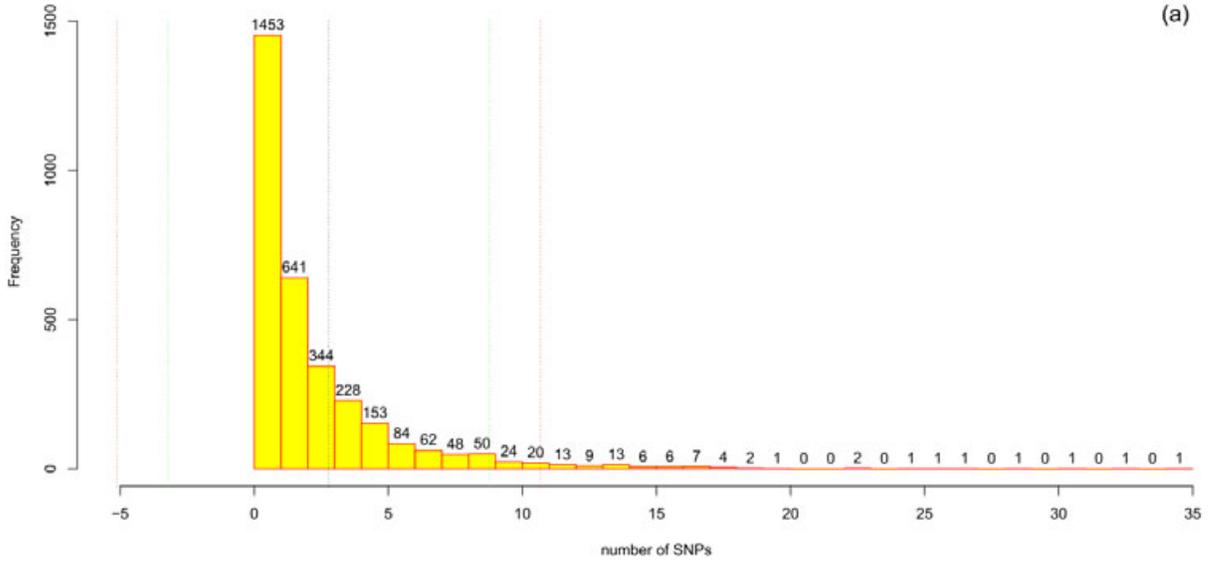


Fig. 3b

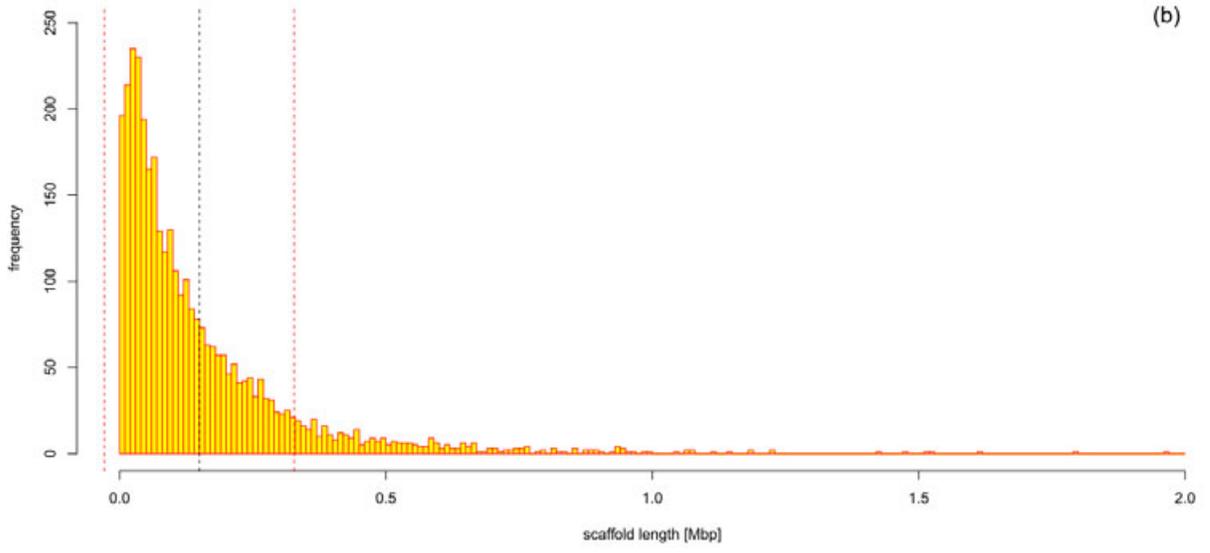


Fig. 3c

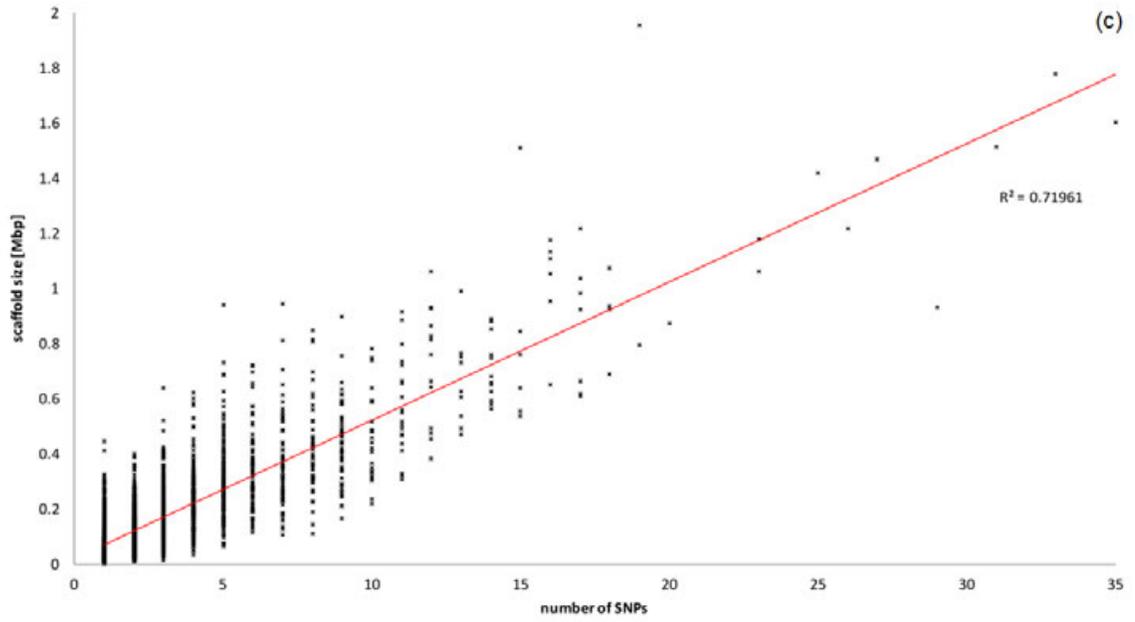


Fig. 4a

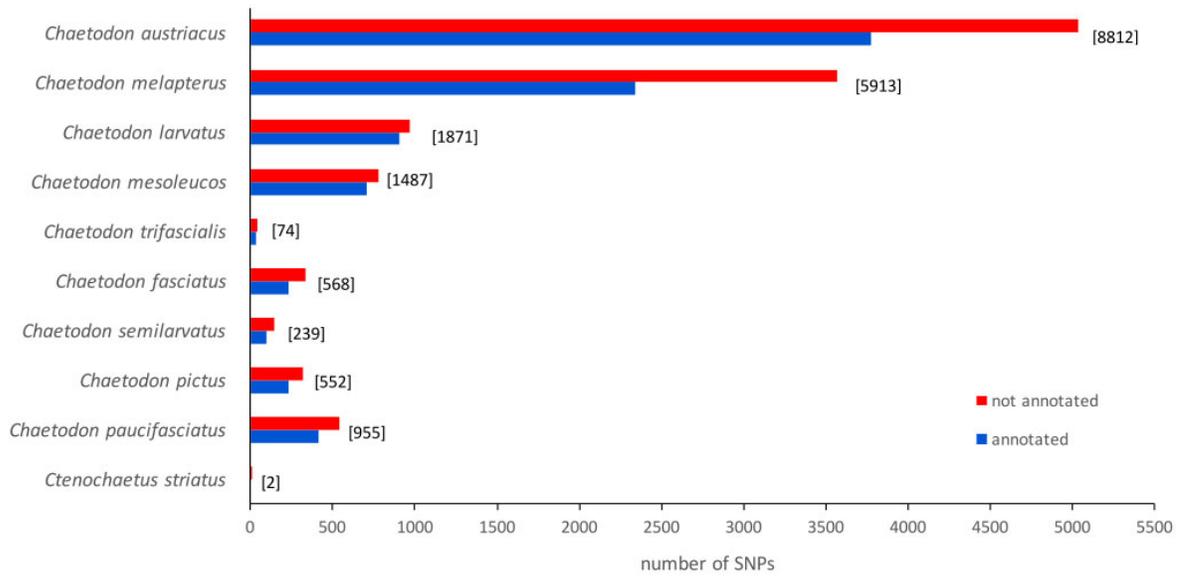


Fig. 4b

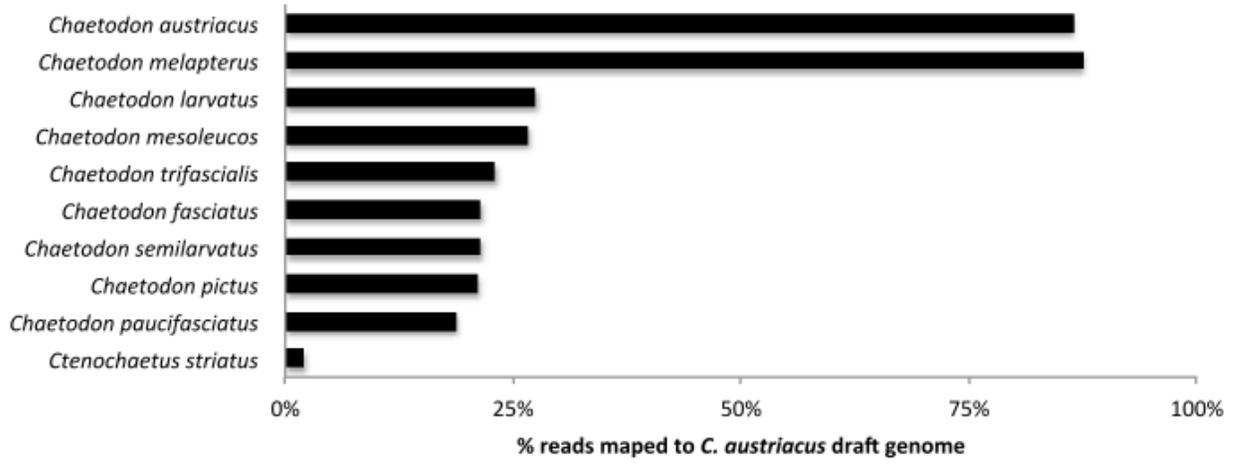


Fig. 4c



Table 1. Stacks results of ddRAD data for nine species of Red Sea and Arabian Sea resident butterflyfish (and an outgroup surgeonfish) before and after quality filtering using *denovo_map.pl* and *ref_map.pl* options in Stacks. In all cases, 12 individuals were sampled per population. Population filtering parameters included: 1) minor allele frequency > 0.05, 2) the locus had to be genotyped in at least 80% of individuals, and 3) the locus was present in all (or all but one) populations, which are represented by numbers outside and inside the parentheses, respectively, for “# of variable loci passing filter”.

Species	Sample size	Number of populations (geographic range of sampling)	denovo_map.pl				ref_map.pl			
			# of reads used	av. depth of coverage	# of loci	# of variable loci passing	# of reads used	av. depth of coverage	# of loci	# of variable loci passing
<i>Chaetodon austriacus</i> (exquisite butterflyfish)	84	7 (Gulf of Aqaba to South Farasan Banks)	106,207,151	13.7	289,504	8,842 (10,711)	91,785,299	13.4	194,235	8,812 (10,780)
<i>Chaetodon fasciatus</i> (Red Sea racoon butterflyfish)	96	8 (Gulf of Aqaba to Djibouti)	93,749,728	10.2	262,590	1,343 (2,650)	20,137,784	12.1	42,029	568 (697)
<i>Chaetodon larvatus</i> (hooded butterflyfish)	60	5 (Thuwal to Djibouti)	85,686,503	12.6	298,347	10,028 (12,393)	23,585,269	13.1	64,641	1,871 (2,179)
<i>Chaetodon melapterus</i> (Arabian butterflyfish)	72	6 (Djibouti to Muscat)	97,774,030	12.2	273,706	2,275 (4,384)	85,591,171	14.9	184,187	5,913 (7,761)
<i>Chaetodon mesoleucos</i> (white-face butterflyfish)	48	4 (Thuwal to Djibouti)	54,334,800	11.9	218,077	7,608 (11,151)	14,493,353	13.3	50,966	1,487 (1,806)
<i>Chaetodon paucifasciatus</i> (Eritrean butterflyfish)	72	6 (Gulf of Aqaba to South Farasan Banks)	100,273,743	16.2	363,221	10,257 (13,539)	18,726,021	16.1	44,073	955 (1,145)

<i>Chaetodon pictus</i> (horseshoe butterflyfish)	60	5 (Djibouti to Masirah Island)	46,939,818	10.9	225,976	2,073 (4,131)	9,907,802	12.8	40,368	552 (759)
<i>Chaetodon semilarvatus</i> (bluecheek butterflyfish)	96	8 (Jazirat Burqan to Djibouti)	89,373,543	12.3	189,554	1,270 (2,053)	19,177,775	13.9	37,766	239 (338)
<i>Chaetodon trifascialis</i> (chevron butterflyfish)	108	9 (Gulf of Aqaba to Masirah Island)	110,578,849	10.7	239,244	476 (1,271)	25,527,940	12.2	45,670	74 (331)
<u>Outgroup</u>										
<i>Ctenochaetus striatus</i> (striated surgeonfish)	120	10 (Gulf of Aqaba to Al Hallaniyats)	126,536,786	9.8	516,163	415 (1,508)	2,573,428	12.4	8,516	8 (27)

Abbreviation: av = average; SNP, single nucleotide polymorphism.