Anacapa Toolkit: an environmental DNA toolkit for processing multilocus metabarcode datasets

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Abstract

1. Environmental DNA (eDNA) metabarcoding is a promising method to monitor species and community diversity that is rapid, affordable, and non-invasive. Longstanding needs of the eDNA community are modular informatics tools, comprehensive and customizable reference databases, flexibility across high-throughput sequencing platforms, fast multilocus metabarcode processing, and accurate taxonomic assignment. As bioinformatics tools continue to improve, addressing each of these demands within a single bioinformatics toolkit is becoming a reality. 2. We present the modular metabarcode sequence toolkit Anacapa (https://github.com/limeybean/Anacapa/), which addresses the above needs, allowing users to build comprehensive reference databases and assign taxonomy to raw multilocus metabarcode sequence data A novel aspect of Anacapa is our database building module, Creating Reference libraries Using eXisting tools (CRUX), which generates comprehensive reference databases for specific user-defined metabarcode loci. The Quality Control and Dereplication module sorts and processes multiple metabarcode loci and processes merged, unmerged and unpaired reads maximizing recovered diversity. Followed by amplicon sequence variants (ASVs) detection using DADA2. The Anacapa Classifier module aligns these ASVs to *CRUX*-generated reference databases using *Bowtie2*. Taxonomy is assigned to ASVs with confidence scores using a Bayesian Lowest Common Ancestor (BLCA) method. The Anacapa Toolkit also includes an R package, ranacapa, for automated results exploration through standard biodiversity statistical analysis.

3. We performed a series of benchmarking tests to verify that the Anacapa Toolkit generates comprehensive reference databases that capture wide taxonomic diversity and that it can assign high-quality taxonomy to both MiSeq-length and Hi-Seq length sequence data. We demonstrate the value of the *Anacapa Toolkit* to assigning taxonomy to eDNA sequences from seawater

samples from southern California including capability of this tool kit to process multilocus metabarcoding data.

4. The *Anacapa Toolkit* broadens the exploration of eDNA and assists in biodiversity assessment and management by generating metabarcode specific databases, processing multilocus data, retaining all read types, and expanding non-traditional eDNA targets. Anacapa software and source code are open and available in a virtual container to ease installation.

1 | Introduction

Rapid and inexpensive biodiversity monitoring tools are critical for maintaining healthy ecosystems and for effective species conservation (Deiner *et al.* 2017). Environmental DNA (eDNA) is a promising non-invasive approach for biodiversity monitoring in terrestrial and aquatic ecosystems because it provides rapid, cost-effective surveys of a broad array of taxa (Taberlet *et al.* 2012; Deiner *et al.* 2017; Bohmann *et al.* 2014; Kelly *et al.* 2014). eDNA metabarcoding is increasingly used in ecology and conservation research, but three key challenges remain in sequence processing and taxonomic assignment stages, limiting the accuracy and reliability of this approach.

To capture broad taxonomic diversity, many eDNA studies simultaneously sequence multiple loci per sample (e.g. Stat *et al.* 2017). However, few metabarcode pipelines are explicitly designed to process multilocus high-throughput sequencing data (but see Arulandhu *et al.* 2017). To process multilocus eDNA sequence data, researchers must sort and process unique metabarcodes independently, which increases valuable user and computation time for each additional metabarcode. bioRxiv preprint doi: https://doi.org/10.1101/488627; this version posted December 7, 2018. The copyright holder for this preprint (which was not certified by peer review) is the author/funder, who has granted bioRxiv a license to display the preprint in perpetuity. It is made available under aCC-BY-NC-ND 4.0 International license.

Another challenge for eDNA metabarcode processing is the lack of robust, locus-specific reference databases (Deiner *et al.* 2017). Curated databases for select metabarcode loci offer validated solutions for certain commonly-used universal metabarcodes (e.g. UNITE Community, 2017), but such curated databases are unlikely to exist for all loci used in metabarcoding studies, especially as the number of target metabarcodes grows. Custom, user-generated databases offer one potential solution, but current approaches can be problematic. For example, generating reference databases through *in silico* PCR will miss reference sequences that do not contain primer recognition sites – which is the case for many sequences available in Genbank (Ficetola *et al.* 2010; Boyer *et al.* 2016). Methods that rely on keyword searches to generate reference databases are sensitive to inaccurate metadata (Machida *et al.* 2017) and are susceptible to retrieving sequences that lack the target metabarcode locus. Together, these issues highlight a need for more comprehensive reference databases that enhance taxonomic assignment for eDNA metabarcoding.

A third challenge of existing metabarcode pipelines is that they frequently discard large portions of sequence data, including reads that can be valuable for assigning taxonomy. For example, many existing pipelines discard unmerged sequences entirely, or only use partial sequence data where full-length alignment with reference metabarcodes is not possible (Port *et al.* 2015). Other tools independently assign taxonomy to unmerged forward and reverse reads assigning the least common ancestor between calls an approach which can lead to poorly-resolved taxonomic assignments (Huson *et al.* 2016). The quality of assignments generated by one pipeline designed to handle unmerged paired data (Bengtsson-Palme *et al.* 2015) suffers from using non-contiguous sequences and suboptimal BLAST parameters (Shah *et al.* 2018). Many pipelines discard significant portions of sequencing data (e.g. remove all reverse reads) for

unmerged paired data, potentially causing selection bias against certain taxa (Deagle *et al.* 2014). Although *k*-mer based classifiers can be modified to accept unmerged paired data, there is no specific documentation or benchmarking for this functionality (e.g. Edgar 2016). Furthermore, the inability of pipelines to incorporate large amounts of sequence information for taxonomic calls will be exacerbated with the use of new high-throughput sequencing platforms (e.g. Illumina NovaSeq and 10X), which generate greater read depth but shorter sequences, which increases the risk of generating unmerged paired data for long metabarcodes.

To help resolve these challenges, we developed the *Anacapa Toolkit*, which contains modules for 1) creating custom reference databases, 2) executing quality control and multilocus read parsing, 3) generating taxonomic assignments for all quality reads produced by HiSeq and MiSeq Illumina platforms, and 4) interactively visualizing taxonomy tables from the *Anacapa Toolkit* using the R package *ranacapa*, described in Kandlikar *et al.* (2018).

2 | Anacapa Toolkit

The *Anacapa Toolkit* combines components of leading bioinformatics software with custom methods (Figure 1). The first module, Creating Reference libraries Using eXisting tools (*CRUX*), generates custom reference databases. The second module performs quality control of raw sequences and infers Amplicon Sequence Variants (ASVs) using *DADA2* (Callahan *et al.* 2016). The third module assigns taxonomy using *Bowtie2* and the Bayesian Lowest Common Ancestor algorithm (*BLCA*; Gao *et al.* 2017).

All components of the *Anacapa Toolkit* are openly available (https://github.com/limeybean/Anacapa; DRYAD link provided upon acceptance). The *Anacapa Toolkit* and several *CRUX*-generated reference databases are available in virtual containers developed with Code for Science and Society (Ogden 2018).

2.1 | *CRUX*: Creating Reference libraries Using eXisting tools

The Anacapa Toolkit's first module, CRUX (Figure 1A; Appendix 1.1), constructs custom reference databases for user-defined primers by querying public databases. CRUX first generates metabarcode-specific seed databases by running *in silico* PCR (Ficetola *et al.* 2010) on the EMBL standard nucleotide database (Stoesser *et al.* 2002). To increase the breadth of reference sequences and capture sequences without barcode primers, CRUX then uses BLASTn (Camacho *et al.* 2009) to query the seed databases against the NCBI non-redundant nucleotide database (Pruitt *et al.* 2005). CRUX de-replicates the BLASTn hits by retaining only the longest version of each sequence and retrieves taxonomy using Entrez-qiime (Baker, 2016). For each primer set, CRUX generates an "unfiltered" database which contains all accessions and taxonomic paths, a paired "filtered" database which excludes accessions with ambiguities in the taxonomic paths, and a Bowtie2-formatted index library (Langmead & Salzberg, 2012).

2.2 | Sequence quality control and ASV parsing

The *Anacapa Toolkit*'s Quality Control and De-replication module (Figure 1B; Appendix 1.2) conducts standard DNA sequence quality control and generates ASVs. It uses *cutadapt* (Martin, 2011) and *FastX-toolkit* (Gordon & Hannon, 2010) to trim user-defined primers and adapters and low-quality bases from raw FASTQ files from Illumina sequencing platforms. Next,

this module uses *cutadapt* to separate reads from multiple loci within each sample. A custom Python script sorts locus-specific reads into three categories: paired-end reads, forward-only reads, and reverse-only reads. These reads are then processed separately through *DADA2* (Callahan *et al.* 2016) to denoise, dereplicate, merge paired reads, and remove chimeric sequences. This step returns ASV FASTA files and ASV count summary tables for four read types: merged paired-end reads, unmerged paired-end reads (filtered based on length and overlap criteria; Appendix 1.2), forward-only reads, and reverse-only reads. These files are inputs for the *Anacapa Toolkit* taxonomic assignment module.

2.3 | Taxonomic assignment: Anacapa Classifier assigns taxonomy with Bowtie2 and BLCA

The Anacapa Classifier module (Figure 1C; Appendix 1.3) assigns taxonomy to ASVs using *Bowtie2* and a modified version of *BLCA* (Gao *et al.* 2017). We verified that our modification to *BLCA* (namely, accepting *Bowtie2*-formatted SAM files rather than BLAST output files) does not influence taxonomy assignment (Appendix 4). In the first step of the Anacapa Classifier, *Bowtie2* queries ASVs against metabarcode-specific *CRUX* generated reference databases returning up to 100 alignments per ASV. The module uses *Bowtie2*'s "very-sensitive" preset to ensure high-quality alignment. The *Bowtie2* outputs are then processed with *Bowtie2-BLCA*, using multiple sequence alignment to probabilistically determine taxonomic identity by selecting the lowest common ancestor from the multiple weighted *Bowtie2* hits for each ASV. This module returns both detailed and brief reports of taxonomic assignment, and eight sets of taxonomy tables based on varying bootstrap confidence cutoffs (40-100).

3 | Benchmarking the Anacapa Toolkit

To benchmark the performance of the first three modules of the *Anacapa Toolkit*, we performed a series of quantitative tests of these modules on various metabarcodes and sequencing read types. We found that *CRUX*-generated databases capture greater taxonomic diversity than published marker-specific reference databases for *CO1*, *12S*, and Fungal *ITS* metabarcodes (Appendix 3). We verified that the Anacapa Classifier consistently generates high-quality taxonomic assignments and explored the consequence of varying bootstrap confidence cutoff on assigned taxonomy (Appendix 4). Finally, we verified that the *Anacapa Toolkit* Quality Control and Classifier modules can process both longer (e.g. MiSeq) and shorter (e.g. HiSeq) DNA metabarcoding sequences (Appendix 5).

4| Case Study : Using the *Anacapa Toolkit* to assign taxonomy to field-collected eDNA samples

To test the *Anacapa Toolkit* on field-collected eDNA metabarcoding datasets, we processed 30 seawater samples from kelp forests across the southern California Channel Islands. Seawater samples were amplified using 12S (Miya *et al.* 2015) and CO1 (Leray *et al.* 2013) metabarcodes (see Appendix 6 for laboratory preparation and data analysis; Table S6.3). For the 12S metabarcode, the *CRUX* module was critical for assigning taxonomy as there are no published reference databases that included all vertebrate species for this locus (Sato *et al.* 2018). Sequence data from these samples are available in NCBI (SRA accession SRP140860). Across seawater samples, we generated 15,745,317 paired-end sequencing reads, of which 11,866,904

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reads and 37,910 ASV were identified as 12S and 3,878,413 reads and 15,997 ASVs were identified as CO1. For both metabarcodes, we found that 99.5% were merged read pairs and <1% were unmerged paired, forward or reverse only reads. The *Anacapa Toolkit*'s taxonomic assignments included 533 assignments at the species level, 414 Genera, 295 Families, 49 Classes, and 21 Phyla of eukaryotes. This included many taxa of interest for natural resource managers (Figure 2; Tables S6.1,S6.2), highlighting the ability of eDNA to detect a wide breadth of marine life and its utility for biodiversity monitoring. A detailed and interactive summary of these seawater samples is available as the demo dataset of the *ranacapa* module (https://gauravsk.shinyapps.io/ranacapa/).

5 | Conclusion

Biodiversity monitoring initiatives are increasingly using eDNA to inventory communities using multilocus metabarcoding. However, a bottleneck in the broad use of eDNA approaches is the lack of bioinformatic pipelines that are accurate and easily customizable. The *Anacapa Toolkit* provides such functionality for eDNA projects as well as other common applications such as gut content analysis (Leray *et al.* 2013) and autonomous reef monitoring structures (Ransome *et al.* 2017). The *Anacapa Toolkit* is modular and its parameters are easily modifiable, so that it can be adapted to users' specific needs in several important ways. First, *CRUX* reference databases are compatible with alternative classifiers (Bokulich *et al.* 2018), and users can append their own reference sequences to *CRUX* databases as needed. Second, the quality control and ASV parsing module are designed to process pooled metabarcoding libraries and automatically sort them by barcode and sample. The resulting output files (with the exception of unmerged reads) can be

analyzed by most classifiers. Third, the *Anacapa Bowtie2-BLCA* Classifier can be applied to any high-throughput sequencing data, and process paired and unpaired reads. The robustness of *CRUX*-generated reference databases and the flexibility of the *Anacapa Toolkit* enables eDNA studies with a variety of metabarcodes to efficiently assign taxonomy in a wide range of community ecology and biodiversity management research.

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Author Contributions

EEC coordinated toolkit development and benchmarking. BS, EEC, GSK, JG, RSM, RW, and ZG designed Anacapa. BS, EEC, GK, JG, MO, and ZG wrote Anacapa. EEC and MO built the Singularity container. EEC, GK, JG, LP, ML, RSM, RW, TO, and ZG designed or conducted benchmarking studies. EEC, LR, TS, and ZG generated the California eDNA libraries. EEC, GK,

and ZG wrote the manuscript with help from LP, LR, ML, RSM and TO; PB, RW, TS revised. NK, PB, and RW provided resources to support the work.

Tables and Figures

Figure 1. Flowchart of the Anacapa Toolkit.

Figure 2. Taxonomic assignments from California environmental samples. **A.** Highlights 12S metabarcodes of Anacapa Island kelp forest vertebrate families. Families in bold are featured.

References

Arulandhu, A.J., Staats, M., Hagelaar, R., Voorhuijzen, M.M., Prins, T.W., Scholtens, I.,
Costessi, A., Duijsings, D., Rechenmann, F., Gaspar, F.B., Crespo, M.T.B., Holst-Jensen, A.,
Birck, M., Burns, M., Haynes, E., Hochegger, R., Klingl, A., Lundberg, L., Natale, C., Niekamp,
H., Perri, E., Barbante, A., Rosec, J.-P., Seyfarth, R., Sovová, T., Moorleghem, C.V., Ruth, S.
van, Peelen, T. & Kok, E. (2017). Development and validation of a multi-locus DNA
metabarcoding method to identify endangered species in complex samples. GigaScience, 6.

Baker, C. (2016). Bakerccm/entrez_qiime: Entrez_qiime v2.0.

Bengtsson-Palme, J., Hartmann, M., Eriksson, K.M., Pal, C., Thorell, K., Larsson, D.G.J. & Nilsson, R.H. (2015). Metaxa2: Improved identification and taxonomic classification of small and large subunit rRNA in metagenomic data. Molecular Ecology Resources, 15, 1403–1414.

Bohmann, K., Evans, A., Gilbert, M.T.P., Carvalho, G.R., Creer, S., Knapp, M., Yu, D.W. & Bruyn, M. de. (2014). Environmental DNA for wildlife biology and biodiversity monitoring. Trends in Ecology & Evolution, 29, 358–367.

Bokulich, N.A., Kaehler, B.D., Rideout, J.R., Dillon, M., Bolyen, E., Knight, R., Huttley, G.A. & Caporaso, J.G. (2018). Optimizing taxonomic classification of marker gene amplicon sequences with QIIME 2's q2-feature-classifier plugin. Microbiome, 6, 90.

Boyer, F., Mercier, C., Bonin, A., Bras, Y.L., Taberlet, P. & Coissac, E. (2016). Obitools: A UNIX-inspired software package for DNA metabarcoding. Molecular Ecology Resources, 16, 176–182.

Budel, B., Dulić, T., Darienko, T., Rybalka, N. & Friedl, T. (2016). Cyanobacteria and algae of biological soil crusts. Biological soil crusts: An organizing principle in drylands, pp. 55–80.
Springer International Publishing.

Callahan, B.J., McMurdie, P.J., Rosen, M.J., Han, A.W., Johnson, A.J.A. & Holmes, S.P. (2016). DADA2: High-resolution sample inference from illumina amplicon data. Nature Methods, 13, 581–583.

Camacho, C., Coulouris, G., Avagyan, V., Ma, N., Papadopoulos, J., Bealer, K. & Madden, T.L. (2009). BLAST+: Architecture and applications. BMC Bioinformatics, 10, 421.

Deagle, B.E., Jarman, S.N., Coissac, E., Pompanon, F. & Taberlet, P. (2014). DNA metabarcoding and the cytochrome c oxidase subunit i marker: Not a perfect match. Biology Letters, 10, 20140562–20140562.

Deiner, K., Bik, H.M., Machler, E., Seymour, M., Lacoursière-Roussel, A., Altermatt, F., Creer, S., Bista, I., Lodge, D.M., Vere, N. de, Pfrender, M.E. & Bernatchez, L. (2017). Environmental

DNA metabarcoding: Transforming how we survey animal and plant communities. Molecular Ecology, 26, 5872–5895.

Edgar, R. (2016). SINTAX: A simple non-bayesian taxonomy classifier for 16S and ITS sequences.

Ficetola, G., Coissac, E., Zundel, S., Riaz, T., Shehzad, W., Bessière, J., Taberlet, P. & Pompanon, F. (2010). An in silico approach for the evaluation of DNA barcodes. BMC Genomics, 11, 434.

Gao, X., Lin, H., Revanna, K. & Dong, Q. (2017). A bayesian taxonomic classification method for 16S rRNA gene sequences with improved species-level accuracy. BMC Bioinformatics, 18.

Gordon, A., & Hannon, G. J. (2010). Fastx-toolkit. FASTQ/A short-reads preprocessing tools (unpublished) http://hannonlab. cshl. edu/fastx_toolkit, 5.

Gu, W., Song, J., Cao, Y., Sun, Q., Yao, H., Wu, Q., Chao, J., Zhou, J., Xue, W. & Duan, J. (2013). Application of the ITS2 region for barcoding medicinal plants of selaginellaceae in pteridophyta. PLoS ONE, 8, e67818.

Huson, D.H., Beier, S., Flade, I., Górska, A., El-Hadidi, M., Mitra, S., Ruscheweyh, H.-J. & Tappu, R. (2016). MEGAN community edition - interactive exploration and analysis of large-scale microbiome sequencing data. PLOS Computational Biology, 12, e1004957.

Kandlikar, G.S., Gold, Z.J., Cowen, M.C., Meyer, R.S., Freise, A.C., Kraft, N.J., Moberg-Parker, J., Sprague, J., Kushner, D.J. & Curd, E.E. (2018). Ranacapa: An r package and shiny web app to

explore environmental DNA data with exploratory statistics and interactive visualizations. F1000Research, 7, 1734.

Kelly, R.P., Port, J.A., Yamahara, K.M. & Crowder, L.B. (2014). Using environmental DNA to census marine fishes in a large mesocosm (G.E. Hofmann, Ed.). PLoS ONE, 9, e86175.

Langmead, B. & Salzberg, S.L. (2012). Fast gapped-read alignment with bowtie 2. Nature Methods, 9, 357–359.

Leray, M., Yang, J.Y., Meyer, C.P., Mills, S.C., Agudelo, N., Ranwez, V., Boehm, J.T. & Machida, R.J. (2013). A new versatile primer set targeting a short fragment of the mitochondrial COI region for metabarcoding metazoan diversity: Application for characterizing coral reef fish gut contents. Frontiers in Zoology, 10, 34.

Machida, R.J., Leray, M., Ho, S.-L. & Knowlton, N. (2017). Metazoan mitochondrial gene sequence reference datasets for taxonomic assignment of environmental samples. Scientific Data, 4, 170027.

Martin, M. (2011). Cutadapt removes adapter sequences from high-throughput sequencing reads. EMBnet journal, 17, 10.

Ogden, Maxwell (2018), CALeDNA Anacapa/CRUX Dat Container (Linux/HPC), v8, UC Merced Dash, Dataset, https://doi.org/10.6071/M31H29

Port, J.A., ODonnell, J.L., Romero-Maraccini, O.C., Leary, P.R., Litvin, S.Y., Nickols, K.J., Yamahara, K.M. & Kelly, R.P. (2015). Assessing vertebrate biodiversity in a kelp forest ecosystem using environmental DNA. Molecular Ecology, 25, 527–541.

Pruitt, K.D. (2005). NCBI reference sequence (RefSeq): A curated non-redundant sequence database of genomes, transcripts and proteins. Nucleic Acids Research, 33, D501–D504.

R Core Team. (2018). R: A language and environment for statistical computing. R Foundation for Statistical Computing, Vienna, Austria.

Ransome, E., Geller, J.B., Timmers, M., Leray, M., Mahardini, A., Sembiring, A., Collins, A.G. & Meyer, C.P. (2017). The importance of standardization for biodiversity comparisons: A case study using autonomous reef monitoring structures (ARMS) and metabarcoding to measure cryptic diversity on Mo'orea coral reefs, French Polynesia (C.A. Chen, Ed.). PLOS ONE, 12, e0175066.

Sato, Y., Miya, M., Fukunaga, T., Sado, T., & Iwasaki, W. (2018). MitoFish and MiFish Pipeline: A Mitochondrial Genome Database of Fish with an Analysis Pipeline for Environmental DNA Metabarcoding. Molecular Biology and Evolution 35, 1553–1555.

Shah, N., Altschul, S.F. & Pop, M. (2018). Outlier detection in BLAST hits. Algorithms for Molecular Biology, 13.

Stat, M., Huggett, M.J., Bernasconi, R., DiBattista, J.D., Berry, T.E., Newman, S.J., Harvey, E.S.
& Bunce, M. (2017). Ecosystem biomonitoring with eDNA: Metabarcoding across the tree of
life in a tropical marine environment. Scientific Reports, 7 .

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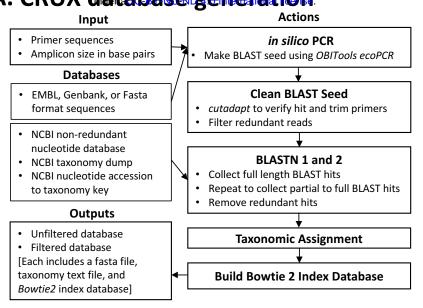
Stoesser, G. (2002). The EMBL nucleotide sequence database. Nucleic Acids Research, 30, 21– 26. 562

Taberlet, P., Coissac, E., Hajibabaei, M. & Riesberg, L.H. (2012). Environmental DNA.

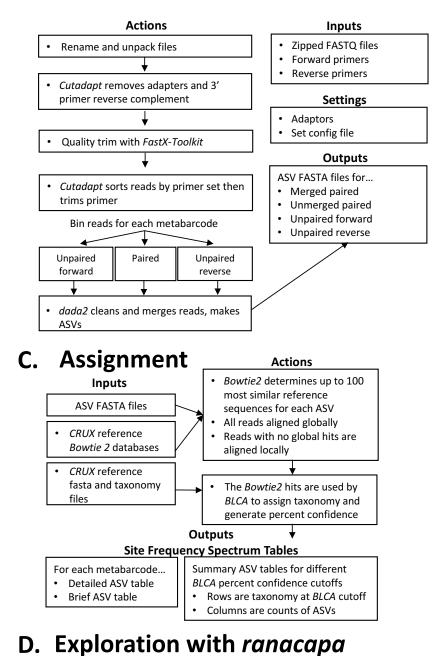
Molecular Ecology, 21, 1789–1793.

UNITE Community. (2017). UNITE qiime release.

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B. Sequence QC and ASV Parsing



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Actinopterygii



Chondritchyes

Mammalia

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	Family	
Acanthuridae	Exocoetidae	Paralichthyidae
Anarhichadidae	Gobiesocidae	Peristediidae
Ascidiidae	Gobiidae	Pholidae
Atherinopsidae	Haemulidae	Phosichthyidae
Balistidae	Hexagrammidae	Pleuronectidae
Batrachoididae	Kuhliidae	Polyprionidae
Blenniidae	Kyphosidae	Pomacentridae
Bythitidae	Labridae	Salmonidae
Carangidae	Labrisomidae	Sciaenidae
Centrarchidae	Lutjanidae	Scombridae
Cichlidae	Malacanthidae	Scorpaenidae
Cirrhitidae	Merlucciidae	Sebastidae
Clinidae	Mullidae	Serranidae
Clupeidae	Muraenidae	Sphyraenidae
Cottidae	Myctophidae	Stichaeidae
Didemnidae	Ophidiidae	Styelidae
Embiotocidae	Opistognathidae	Syngnathidae
Engraulidae	Oplegnathidae	Zaniolepididae

Laridae Phalacrocoracidae

Sulidae

Alopiidae	
Arhynchobatidae	
Dasyatidae	
Myliobatidae	
Rajidae	
Scyliorhinidae	
Squatinidae	
Torpedinidae	
Triakidae	
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Delphinidae Otariidae Phocidae