

1 CS 5970: Bioinformatics Project

2 Genomic Basis of Exaggerated Dorsal Fin Morphology in Livebearing Fishes

3 Tyler J. Reich

4 **Abstract:**

5 Morphological diversity in closely related species provides powerful opportunities to
6 understand how genetic variation shapes phenotypic evolution. In livebearing fishes of the genus
7 *Poecilia*, dorsal fin length varies widely across species, with some exhibiting “sailfin”
8 morphologies and others possessing short, rounded fins. Hybrids between long- and short-finned
9 species display intermediate phenotypes, suggesting a quantitative genetic basis for this trait.
10 Despite its ecological and evolutionary significance, the genetic architecture underlying dorsal
11 fin length remains unknown.

12 This project aims to develop and implement a bioinformatics workflow to identify
13 genomic regions associated with dorsal fin length variation by integrating low-coverage whole-
14 genome sequencing (lcWGS) with quantitative trait loci (QTL) mapping in hybrid populations. I
15 have already generated F1 hybrids between *Poecilia latipinna* (Sailfin molly) and *P. mexicana*
16 (Atlantic molly) and am producing F2 hybrids for genomic analysis. Building on approaches
17 established by Powell et al., (2021), this project will focus on designing and testing a complete
18 computational pipeline for processing lcWGS data, performing variant calling, generating
19 ancestry-informative markers, and conducting QTL scans to identify chromosomes and genomic
20 intervals influencing dorsal fin length.

21 The outcome of this work will be a reproducible, well-documented analysis pipeline
22 ready for deployment once sequencing data become available. This project will directly
23 contribute to understanding genotype-phenotype relationships in *Poecilia*, inform future
24 transcriptomic analyses of fin development, and support my dissertation research on the
25 evolution of morphological variation in livebearing fishes.

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41 **Specific Aims:**

42 Variation in dorsal fin morphology across species of *Poecilia* represents an evolutionarily
43 important trait whose genetic basis remains unknown. Long-finned species (e.g., *Poecilia*
44 *latipinna*) and short-finned species (e.g., *P. mexicana*) differ markedly in dorsal fin length; and
45 hybrids show intermediate phenotypes, indicating a quantitative genetic basis. Identifying the
46 genomic regions responsible for this variation will provide crucial insight into how
47 morphological traits evolve and diversify in livebearing fishes.

48 The goal of this project is to develop and implement a complete computational workflow
49 for analyzing low-coverage whole-genome sequencing (lcWGS) data from F2 hybrids to identify
50 quantitative trait loci (QTL) associated with dorsal fin length. This work builds directly on
51 established methods used in the closely related *Xiphophorus* and *Poecilia* hybrid systems and
52 lays the foundation for the genomic analyses that will form a central component of my
53 dissertation research.

54 **Aim I: Develop a reproducible pipeline for processing low-coverage whole-genome
55 sequencing data and generating high-quality genotype likelihoods.**

56 I will implement a bioinformatics workflow for trimming, aligning, and filtering lcWGS
57 data from F2 hybrids generated by crossing *P. latipinna* and *P. mexicana*, then performing F1
58 hybrid intercrosses. This aim includes identifying ancestry-informative markers (AIMs),
59 performing variant calling using genotype-likelihood-based approaches (e.g., GATK), and
60 generating the input files required for QTL analysis. Innovation: Applying low-coverage,
61 likelihood-based genomic methods to this system reduces sequencing cost while retaining power
62 for detection. Outcome: A validated, modular pipeline for lcWGS processing and variant
63 calling.

64 **Aim II: Perform quantitative trait locus (QTL) mapping to identify genomic regions
65 associated with dorsal fin length.**

66 Using phenotypic measurements from F2 hybrids and the genotypic data generated in
67 Aim I, I will conduct QTL mapping to detect chromosomes and genomic intervals underlying
68 dorsal fin length variation. This includes model selection, permutation testing, and visualization
69 of QTL effect sizes. Innovation: This will represent the first QTL analysis of dorsal fin
70 morphology in *Poecilia* using F2 hybrids and establishes a novel framework for mapping
71 morphological traits in this group. Outcome: Identification of candidate chromosomes and
72 genomic regions influencing fin length and a foundation for future fine-mapping and gene
73 expression analyses.

74 **Impact:**

75 Completion of these aims will produce a fully documented, analysis-ready workflow for
76 lcWGS-based QTL mapping and generate genomic insights into the genetic architecture of
77 dorsal fin variation in mollies. This work will directly support downstream RNA-seq studies of
78 candidate genes, strengthen the genomic tools available for *Poecilia*, and accelerate progress
79 toward understanding how complex morphological traits evolve.

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85 **Research Strategy:**

86 Significance:

87 Exaggerated morphological traits provide
88 powerful systems for uncovering how genetic variation
89 gives rise to phenotypic diversity, yet the genetic
90 architecture of many sexually selected ornaments remains
91 poorly understood. In livebearing fishes of the genus
92 Poecilia, dorsal fin morphology varies dramatically
93 among species, with Sailfin mollies (*P. latipinna*)
94 exhibiting a hypertrophied, sexually selected fin while
95 closely related species such as *P. mexicana* retain short,
96 unornamented fins (Figure 1) (Farr et al., 1986; Reznick
97 et al., 2017; Goldberg et al., 2019; Reznick et al., 2021).
98 Hybridization between these species produces
99 intermediate dorsal fin lengths, strongly suggesting a
100 quantitative genetic basis for this trait and highlighting
101 the system as an ideal model for dissecting genotype-
102 phenotype relationships (Berbel-Filho et al., in prep).



Figure 1: Males of *Poecilia latipinna* (top) and *P. mexicana* (bottom) exhibiting drastic dorsal fin morphology, particularly in length.

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104 Despite this promise, only one study has attempted to identify genomic regions
105 influencing dorsal fin exaggeration in *Poecilia* (Keong et al., 2014), and that work relied solely
106 on a small microsatellite (SSR) panel of 29 loci – only 18 of which were informative and
107 grouped into four linkage groups – rather than whole-genome sequencing, limiting both genomic
108 resolution and power to detect polygenic effects. As a result, the field still lacks fundamental
109 knowledge of: (1) how many loci contribute to dorsal fin length variation, (2) whether the trait is
110 governed by a few large-effect regions or a polygenic architecture, and (3) whether the genetic
111 basis of fin exaggeration overlaps with known sexually selected ornaments in other poeciliids,
112 such as swordtails (*Xiphophorus*) (Powell et al., 2021).

113 This lack of genomic resolution represents a major barrier to progress. Without high-
114 density markers and genome-wide scans, the evolutionary and developmental mechanisms
115 underlying dorsal fin exaggeration remain unresolved, limiting our ability to connect phenotypic
116 evolution to underlying genetic processes. By applying low-coverage whole-genome sequencing
117 (lcWGS) and QTL mapping in F2 hybrids, this project directly addresses this gap and will
118 establish the first high-resolution genomic framework for understanding the evolution of
119 exaggerated fin morphology in *Poecilia*.

120 Innovation:

121 This project introduces several methodological and conceptual innovations that advance
122 the study of morphological evolution in *Poecilia* and extend current research paradigms in fish
123 genomics.

124 **1. First application of whole-genome-based QTL mapping for dorsal fin morphology in**
125 ***Poecilia*:**

126 Previous work on dorsal fin exaggeration relied exclusively on low-resolution
127 microsatellite markers (Keong et al., 2014), which limited the ability to detect small-effect loci or
128 define precise genomic intervals. By leveraging low-coverage whole-genome sequencing
129 (lcWGS), this project provides the first high-density, genome-wide characterization of the
130 genetic architecture underlying dorsal fin length in mollies.

131 **2. Implementation of a cost-efficient, likelihood-based genomic workflow:**

132 The project applies cutting-edge lcWGS pipelines that use genotype likelihoods rather than
133 hard-called SNPs, allowing accurate QTL detection even at low sequencing depth. This
134 approach, recently adopted in other teleost hybrid systems (Powell et al., 2021), substantially
135 reduces sequencing costs while preserving statistical power, making it an innovative model for
136 morphological mapping in non-model fishes.

137 **3. Use of F2 hybrids to dissect quantitative trait architecture across species boundaries:**

138 By mapping dorsal fin variation in F2 hybrids between *P. latipinna* and *P. mexicana*, this
139 project captures recombination across divergent genomes, enabling fine-scale detection of loci
140 influencing fin morphology. This design provides stronger resolution than within-species
141 mapping and represents a novel comparative genetic framework for *Poecilia* ornamentation.

142 **4. Integration with downstream functional genomics:**

143 The pipeline developed here is explicitly designed to interface with future RNA-seq and
144 candidate gene expression analyses of fin development. This forward-looking integration
145 distinguishes the project from earlier work that ended at marker identification (Keong et al.,
146 2014), opening the door to mechanistic studies of the developmental pathways shaping
147 exaggerated fin structures.

148 **5. Creation of a generalizable, reproducible pipeline through collaboration with Dan
149 Powell:**

150 This project leverages an active collaboration with Dr. Dan Powell, whose team developed a
151 robust lcWGS and QTL-mapping pipeline for identifying the genetic architecture of sexually
152 selected fin ornaments in *Xiphophorus* (Powell et al., 2021). By adapting and extending this
153 proven workflow to analyze dorsal fin length in *Poecilia*, this project transforms an ornament-
154 mapping framework originally designed for swordtail fish into a generalizable, cross-taxon toolset
155 for studying morphological evolution. The resulting pipeline – documented, modular, and
156 compatible with low-coverage genomic data – will provide a reusable resource not only for the
157 *Poecilia* community but also for researchers working on hybrid systems, sexual selection, and
158 trait evolution in other fish lineages. This represents the first systematic translation of the
159 *Xiphophorus* ornament-mapping pipeline to another genus, demonstrating its broader
160 applicability and significantly expanding the genomic toolkit available for poeciliid evo-devo
161 research.

162 Together, these innovations modernize genetic mapping in *Poecilia*, expand the toolkit for
163 investigating sexually selected ornamentation, and establish a new methodological foundation
164 for linking genotype to phenotype in livebearing fishes. By integrating low-coverage whole-
165 genome sequencing, genotype-likelihood-based variant calling, and hybrid-mapping designs, this
166 project brings Poecilia genetics in line with contemporary standards used in model systems. The
167 resulting pipeline not only increases power and resolution relative to earlier microsatellite-based
168 approaches (Keong et al., 2014) but also creates a scalable framework that can incorporate future
169 datasets, including RNA-seq and long-read assemblies. In doing so, this work positions *Poecilia*
170 as a tractable genomic system for dissecting the molecular basis of complex traits, enabling
171 downstream research into regulatory evolution, convergent genetic pathways with *Xiphophorus*
172 (Powell et al., 2021), and the broader mechanisms by which sexual selection shapes
173 morphological diversity.

174 Approach:

175 To generate the mapping population, *Poecilia latipinna* (sailfin) will be crossed with *P.*
176 *mexicana* (shortfin/Atlantic molly) to produce F1 hybrids, which will then be intercrossed to
177 generate an F2 population segregating for dorsal fin length (Figure 2) (Lander & Botstein, 1989).
178 A sufficiently large F2 cohort (target n = 200-300) will be maintained to ensure adequate
179 statistical power for detecting QTL, including small-effect loci in a potentially polygenic

180 architecture. All F2 individuals will be phenotyped
181 for dorsal fin length, height, and overall shape
182 using standardized digital imaging and
183 morphological analysis such as ImageJ (Schneider
184 et al., 2012). Additional covariates, including total
185 body length and sex, will be recorded to account
186 for allometric scaling and sexual dimorphism in
187 downstream QTL analyses.

188 Genomic DNA will be extracted from all F2
189 individuals and used to prepare sequencing libraries
190 for low-coverage whole-genome sequencing
191 (lcWGS) at approximately 1x coverage per
192 individual (Li et al., 2009). Reads will be aligned to
193 both parental reference genomes (*P. latipinna* and
194 *P. mexicana*) using Burrows-Wheeler Aligner
195 (BWA) (Li & Durbin, 2009) and processed with
196 Sequence Alignment/Map (SAM) tools (Li et al.,
197 2009) to maximize accuracy in ancestry assignment
198 and variant detection. This dual-reference
199 alignment improves assignment of ancestry at each
200 locus, enhances detection of species-specific
201 alleles, and minimizes mapping bias that can occur
202 when using a single reference, thereby increasing
203 the power and precision of downstream QTL
204 analyses. Standard quality-filtering procedures will
205 be applied to remove low-quality or ambiguous
206 reads, ensuring reliable results while maintaining
207 cost efficiency for the large F2 population.

208 Genotype calling will be performed using Genome Analysis Toolkit (GATK)'s genotype-
209 likelihood-based methods (McKenna et al., 2010), allowing robust Single Nucleotide
210 Polymorphism (SNP) identification from low-coverage sequencing data. Ancestry-informative
211 markers (AIMs) that distinguish *P. latipinna* and *P. mexicana* alleles will be generated and local
212 ancestry of F2s will be inferred using a Hidden Markov Model (HMM) framework (Liu et al.,
213 2014; Powell et al., 2021). The resulting genotype data will be combined with phenotypic
214 measurements of dorsal fin traits to produce the input files necessary for QTL mapping. This
215 approach leverages GATK's established framework, ensuring accurate variant calling while
216 maintaining compatibility with downstream analyses.

217 Genome-wide QTL scans will be conducted using R/qtl to identify loci associated with
218 dorsal fin traits (Broman et al., 2003). Significance thresholds will be determined through
219 permutation testing. Confidence intervals for detected QTL will be estimated to assess the
220 precision of each signal. Effect sizes and potential interactions among loci will be modeled to
221 distinguish major-effect loci from polygenic contributions, providing insight into the genetic
222 architecture of dorsal fin exaggeration. Identified QTL will be compared with candidate regions
223 previously reported in *Xiphophorus* (Powell et al., 2021) to investigate potential convergent
224 genetic mechanisms underlying sexually selected fin traits across poeciliids. These loci will help
225 reveal how sexually selected traits are encoded genetically and how they may evolve under
226 selection.

227 Several challenges may arise during this project, along with strategies to mitigate their
228 impact. Low sequencing coverage could reduce the accuracy of rare allele detection; to address
229 this, coverage can be increased for a subset of individuals to validate critical loci, or imputation
230 strategies leveraging parental genotypes can be applied. The potentially complex or polygenic

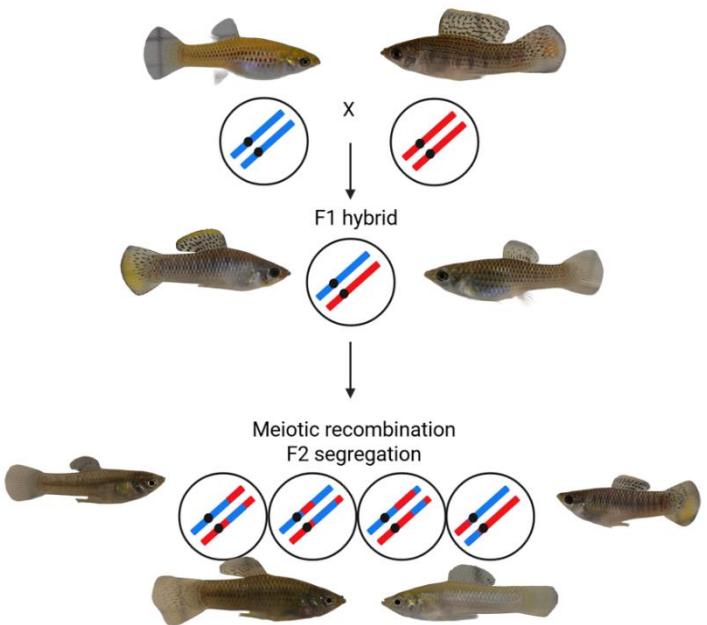


Figure 2: Schematic demonstrating F2 intercross design between a *Poecilia latipinna* male (top right) and *P. mexicana* female (top left) with F2s segregating for dorsal fin size (bottom).

231 architecture of dorsal fin traits may reduce power to detect individual QTL; multi-locus or
232 Bayesian mapping approaches can be employed, and the F2 sample size can be increased if
233 initial scans suggest widespread polygenic effects. Finally, environmental variation may
234 influence phenotypic measurements, so rearing conditions will be standardized to control for
235 non-genetic sources of variation.

236 **Results:**

237 To establish a robust pipeline for processing low-coverage whole-genome sequencing
238 (lcWGS) data and generating high-quality genotype likelihoods, I first validated the workflow
239 using human genomic data as a proof-of-concept. Reference genomes were indexed using
240 Burrows-Wheeler Aligner (BWA) (Li & Durbin, 2009) and paired-end reads were aligned using
241 the BWA-MEM algorithm (which identifies Maximal Exact Matches), producing Binary
242 Alignment/Map (BAM) files (Li, 2013). These BAM files were subsequently sorted, duplicates
243 marked, read groups added, and indexed using the Sequence Alignment/Map (SAM) toolkit (Li
244 et al., 2009) to ensure compatibility with downstream variant calling. Variant calling was
245 performed using the Genome Analysis Toolkit (GATK; McKenna et al., 2010; DePristo et al.,
246 2011) with the HaplotypeCaller algorithm run in reference-confidence mode to produce genomic
247 Variant Call Format (gVCF) files, which contain per-site genotype likelihoods. These single-
248 sample gVCFs were subsequently genotyped into standard Variant Call Format (VCF) files. This
249 process successfully traversed multiple chromosomes, confirming the pipeline's ability to handle
250 large eukaryotic genomes. Test runs verified that all components – BWA, SAMtools, and GATK
251 – function correctly in a WSL2 Linux environment on a Windows workstation, and outputs
252 matched expected formats, demonstrating reproducibility and reliability.

253 To validate the QTL mapping portion of the pipeline, I used the *listeria* dataset, an
254 example F2 intercross dataset provided in R/qltl (Broman et al., 2003). Using this dataset, I
255 confirmed that genotypes and phenotypes could be correctly imported and inspected, missing
256 data visualized, and segregation ratios checked. I then calculated genotype probabilities,
257 performed genome scans using Haley-Knott regression (Haley & Knott, 1992), and conducted
258 permutation tests to determine genome-wide significance thresholds. For significant loci, I
259 estimated 95% confidence intervals and visualized marker effects, including the position of peak
260 markers within confidence intervals (Figures 3-5). This analysis demonstrates that the workflow
261 is fully operational for F2 intercross data and can be directly applied to my *Poecilia* sequencing
262 data once it becomes available.

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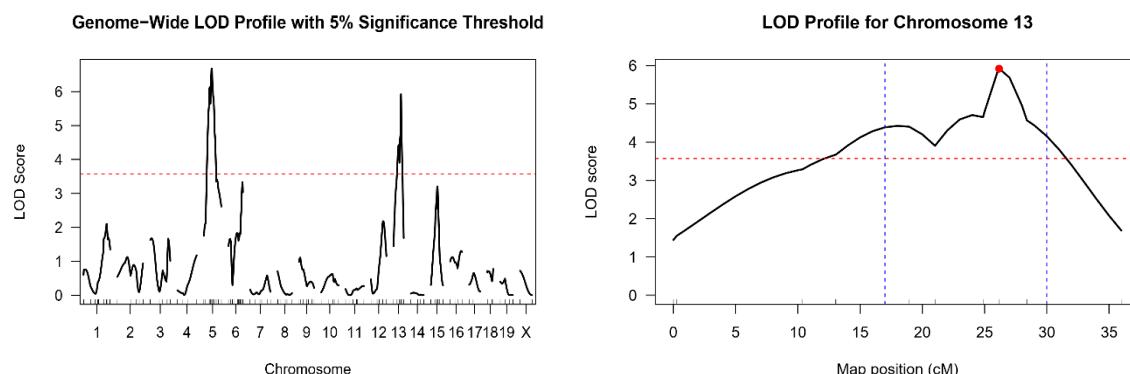
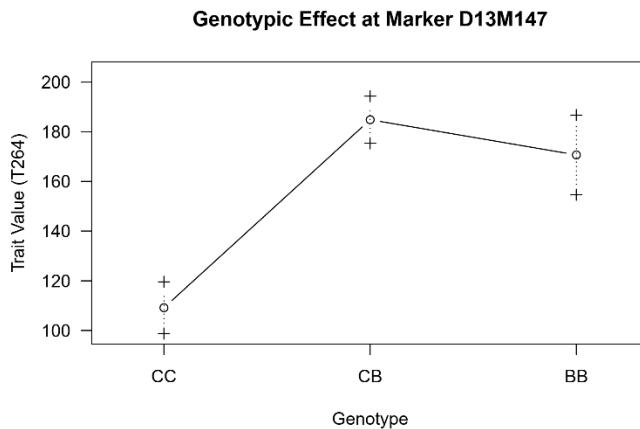


Figure 3: The genome-wide QTL scan with the permutation-derived 5% significance threshold overlaid (red horizontal line). This figure highlights genomic regions indicating statistically significant QTL (5 & 13).

Figure 4: Chromosome-specific LOD profile for chromosome 13 including confidence interval boundaries (blue vertical lines) and the peak marker (red circle) associated with the strongest signal, D13M147.

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Figure 5: Effect of genotype classes at peak marker (D13M147) on the phenotype (T264). This plot visualizes how phenotypic values differ among genotypes, providing insight into the direction and magnitude of the QTL effect.

276 With the pipeline established and validated, application to my F2 *Poecilia* sequencing
277 data will require updating only the reference genomes, input FASTQ filenames, and phenotype
278 data. Once these inputs are available, the workflow can efficiently generate high-quality
279 genotype likelihoods, enable accurate inference of ancestry-informative markers, and prepare
280 input files for genome-wide QTL mapping of dorsal fin traits. The full workflow, including
281 scripts and example data, is available on [my GitHub page](#).

282 Future Directions:

283 If this project is expanded into a full research program with publication goals, several
284 avenues could be pursued. First, candidate loci identified through QTL mapping could be
285 validated and functionally characterized using RNA-seq in developing dorsal fins of F2 hybrids
286 and parental species, linking genotype to gene expression and developmental mechanisms.
287 Second, high-priority loci could be further investigated with targeted functional assays,
288 leveraging existing poeciliid genetic tools, to directly test their contributions to dorsal fin
289 exaggeration and sexually selected ornamentation. Finally, comparative analyses across other
290 poeciliid species or independent ornamental traits (e.g., *Xiphophorus* swordtails) could explore
291 whether similar genetic and developmental pathways underlie convergent evolution of fin
292 exaggeration, providing broader insight into the evolution of sexually selected traits in
293 livebearing fishes.

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