

Supplementary Information for

Unraveling the genotype by environment interaction in a thermosensitive fish with a polygenic sex determination system

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Supplementary Information Text

Supplementary Materials and Methods

Genotyping

SNP calling was done using the ThermoFisher AxiomAnalysisSuite software (v5.1.1.1, Affymetrix, Santa Cara, CA) and quality controls with PLINK 1.9 (Purcell and Chang, 2015).

A first quality control of the genotypes excluded 79 individuals having a genotyping rate lower than 90%: 24 larvae sampled at 77 DD_{10 °C}, 53 larvae sampled at 242 DD₁₀ °c, and 2 fish sexed at one year. Using the remaining genomic data set, we applied classical quality control by removing SNPs with a minor allele frequency (MAF) inferior to 5%. The resulting data set included 49,543 SNPs on the parents, grandparents, and the remaining 1951 offspring. Using a subset of 1,000 highly polymorphic SNPs, we reconstructed the pedigree of the offspring using the R package APIS (1). The number of fish assigned to a single parental pair was 1929 out of the 1951 genotyped (98.9%).

eGST, QTL presence and heritability assessment

For this estimation, we only kept the offspring with a sex phenotype, a genotype and a pedigree, for a total of 1107 individuals, from now on referred to as the "training" dataset. The following model was used to fit the sex phenotype:

 $y = Xb + Zu + e$ (Model 1)

where y is the vector of binary sex phenotypes ($1 =$ male and $2 =$ female), b is a vector of the fixed effects (intercept and temperature). **X** is the incidence matrix relating phenotypes with the fixed effects. **Z is** the incidence matrix relating phenotypes with the random animal genetic effects. *u* is the vector of random animal genetic effects with the following distribution N (0, $\mathbf{G}\sigma^2\mathbf{g}$), where **G** is the genomic relationship matrix, σ²g the additive genetic variance, and **e** a vector of residuals following a *N*(0,**I**σ2e) distribution, where σ2e is the residual variance and **I** the identity matrix. Model 1 was applied for sex as a single trait (considering intercept and temperature as fixed effects). A multi-trait animal model (Model 2) with sex considered a different trait at LT and at HT (sex_LT and sex HT), with intercept as the only fixed effect, was used to estimate the genetic correlation between sex_LT and sex_HT.

The variance components (σ ²g and σ ²e) were estimated using a Gibbs sampler with 500,000 iterations, 100,000 burn-in, and one sample was kept every 20 iterations for posterior analysis. The residual variance σ^2 e was set to a value of 1. The posterior distributions were analyzed with the R package boa (2).

The heritability of sex on the PST scale was estimated as:

$$
h_u^2 = \frac{\sigma^2 g}{\sigma^2 g + \sigma^2 e}, \quad (2)
$$

Using the sex single-trait threshold model (Model 1) built from the training dataset, and the Genomic-based Estimated Breeding Values (GEBV) of PST, we could obtain an estimated Genetic Sex Tendency (eGST) for the rest of the offspring dataset (offspring with a genotype and a pedigree, but no sex phenotype N= 844) by using the POSTGIBBSF90 software. The accuracy of the model was assessed by applying a leave-one-out cross-validation method on the training population. For this cross-validation, the phenotype of individual *i* was masked, and the resulting threshold model was used to predict its eGST*i*, which was then compared to its actual sex phenotype *yi*, using the Receiver Operating Characteristic curves (ROC curves) methodology to assess the efficacy of classifying the animals as male or female. The Area Under Curve (AUC) metric (3) was used to interpret the performance of the genomic prediction models, with values of 1 representing the perfect classifier.

Following the same model as described above (Model 1), we used the posterior SNP solutions to apply a weighted genome wide association scan (wGWAS)(4). At the first iteration, **G** was calculated as **G** = **ZDZ′**/*q*, with **D** equal to the identity matrix (5). Then, direct genomic values (DGVs) were obtained from GEBVs as DGVi = −(Σ*j*, *j*≠*ⁱ gij*GEBV*^j* /*gii*) with *gij* elements of the **G**−1 matrix (6), and converted to SNP effects as $a_i = DZ'G^{-1}DGV_i$. A new **D** matrix was computed from the genetic variance explained by each SNP. After two rounds, no further modification of the variance explained by the SNPs was seen. The LOD scores of SNPs were computed as -log(*pvalue*) and a regional variance explained was computed by summing variance explained in 50 SNPs sliding windows. We considered as QTLs the genomic segments that explained a proportion of genetic variance higher than 2%, as done in (7).

RNA extraction, RNA-seq and RNA-seq data analysis

RNA extraction

The gonads of juveniles were particularly small, and the first trial of mRNA extraction with a dedicated kit failed (n = 27). We thus performed mRNA isolation using QIAzol lysis reagent (Quiagen) on remaining gonad samples (n = 43, 22 for LT and 21 for HT) and on individuals at the "flexion" and "all fins" stages. All samples were mechanically disrupted using ball mills using the appropriate QIAzol quantity, following manufacturer instructions. The RNA concentration was assessed by measuring the A260/A280 ratio using the NanoDrop® ND-1000 V3300 Spectrophotometer (Nanodrop Technology Inc., Wilmington, DE, USA) and RNA quality was checked using an Agilent bioanalyzer (Agilent Technologies, Santa Clara, CA, USA) via the

electrophoretic trace method. The RNA integrity number was between 8.5 and 10 for all samples (Table S4). One extraction failed at the "all fins" stages, and one sample presented the lowest RNA quantity, and was removed from the analysis (n total $= 68$).

RNA-Seq

All RNAs available at the "all fins" stage ($n = 68$) and at the juvenile stage ($n = 43$) were used for RNA-Seq. In addition, because of spare space on the sequencer, RNA-Seq was also performed on 10 individuals at the "flexion" stage. We selected five individuals from the LT group and five from the HT group based on their eGST, choosing those which were genetically "weak" females (i.e. with a positive, but not extreme eGST: 0 < eGST < 0.5).

Because of the difference in RNA quantities extracted from whole bodies or from gonads, cDNA library construction was performed using two separate methods. For whole bodies, libraries construction was described in (8) at the "all fins" stages and was performed similarly at the "flexion" stage.

cDNA library construction of gonads at the juvenile stage

The 43 libraries from juvenile gonad samples were prepared using a NuGEN Universal Plus mRNA-Seq kit (Tecan, Redwood city, CA, USA) according to the manufacturer's instructions. Briefly, polyadenylated RNAs were selected on oligo(dT) magnetic beads. Chemical fragmentation of selected RNAs was done. The first-strand cDNA synthesis started (random primers) in the presence of Actinomycin D, in order to prevent spurious DNA-dependent synthesis, improving strand specificity. Following this, the second strand cDNA synthesis and repairing the ends of double-stranded cDNA fragments followed. Then, adapters were ligated at both ends of fragmented nucleic acid, preparing them for hybridization onto a flow cell. Next, a selection of the strand to be sequenced and purification were done. Ligated cDNAs were amplified following a 15-cycle PCR. This step also leads to a selection, since only fragments at both ends will be enriched. PCR products were purified using AMPure XP Beads (Beckman Coulter Genomics, Brea, CA, USA). Libraries were validated using a Fragment Analyzer (Agilent Technologies, Santa Clara, CA, USA) and quantified using the KAPA Library quantification kit (Roche, Bâle, CHE).

Libraries sequencing:

The 10 "flexion", 68 "all fins" and 43 gonad samples libraries were pooled in equimolar amounts, denatured with NaOH, and diluted to 20-22 pM before the clustering. This was performed on a cBot system (Illumina, San Diego, CA, USA) and sequencing was performed on an HiSeq 2500 (Illumina, San Diego, CA, USA) using the single-end 1*50 nt protocol on 9 lanes of a flow cell V4 or V2. Sequencing produced between 5 and 34 million passed filter clusters per library (9).

Sequencing quality control

Image analyses and base calling were performed using the Illumina HiSeq Control Software and the Real-Time Analysis component. Demultiplexing was performed using Illumina's conversion software (bcl2fastq 2.20). The quality of the raw data was assessed using FastQC from the Babraham Institute and the Illumina software SAV (Sequencing Analysis Viewer). Potential contaminants were monitored with the FastQ Screen software from the Babraham Institute.

RNA-Seq data analysis

A splice junction mapper, TopHat 2.1.1 (10) using Bowtie 2.3.4.3 (11), was utilized to align the RNA-Seq reads to the *Dicentrarchus labrax* genome (NCBI, reference GCA_000689215.1) with a set of gene model annotations (12). Final read alignments having more than three mismatches were discarded. Samtools (v.1.9)(13) was used to sort the alignment files. Then, the counting was performed with Featurecounts (v.1.6.4 (14)). For both the "flexion" and the "all fins" samples (libraries with high starting material), reads were mapped to the opposite strand of the gene (-s2 option), while for gonad samples (libraries with low starting material), the reads were mapped to the strand of the gene (-s1 option). Before statistical analyses, genes with less than 30 reads (mean of all the analyzed samples) were filtered out.

RRBS sequencing and preliminary analysis

RRBS pools were sequenced on a HiSeq2500 (Illumina, San Diego, CA) using 50 base pair (bp) single-read sequencing to obtain at least 22.2 million reads/sample (~15x coverage). The adaptertrimmed sequence reads were aligned to the reference genome (seabass_V1.0.104, GCA_000689215.1) (12), and methylation values were extracted using Bismark v0.20.0 (15) with default parameters. Following, MethylKit v1.18.0 (16) was used to remove data with low coverage $\left($ <10x) and data with extremely high coverage $\left($ >99.9th percentile of coverage in each sample) to discard reads with possible PCR bias. Furthermore, data were normalized for read coverage distribution between samples.

Data Analysis

Gene Ontology selection

We selected several GO biological processes based on our hypotheses: Sex determination (GO:0007530), Sex differentiation (GO:0007548), Steroids metabolic process (GO:0008202); Regulation of growth rate (GO:0040009), Lipid biosynthetic process (GO:0008610), Execution phase of apoptosis (GO:0097194); Glucocorticoid metabolic process (GO:0008211); Cellular response to heat (GO:0034605); Regulation of gene expression, Epigenetic (GO:0040029); Histone modification (GO:0016570); DNA methylation or demethylation (GO:0044728); Gene silencing by miRNA (GO:0035195); Gene silencing by RNA (GO:0031047); Regulatory RNA binding (GO:0061980) and Germ cell development (GO:0007281). The presence of the genes in these GOs was then tested in regards to the list of genes previously identified as significantly expressed in the models described in the M&M of the main file.

Fig. S1. Scheme of the gonadal development of the European Sea bass, modified from (17). The dotted black arrows indicate the size at which larvae were sampled for molecular analyses. The first sampling is not represented since no size (in length) is available and that these larvae only served for genotyping. The blue and red arrows indicate at which age LT and HT larvae were collected, respectively. Red triangle means masculinization when individuals are exposed too early to high temperature based on age at the start of experiment (17), and blue triangle means masculinization when animals are exposed to low temperature for a relatively long period (based on age)(18).

Fig. S2. Receiver operator characteristic (ROC) curve and corresponding area under the curve (AUC) metric for genetic sex tendency (eGST) of European sea bass.

Fig. S3. Density plot of the genetic sex tendency (eGST) of European sea bass at four distinct developmental stages and two temperatures: HT= high temperature 21 °C; LT= low temperature 16 °C. The number (n) of genotyped individuals is indicated. The blue (LT) and red (HT) dotted lines indicate the mean genetic sex tendency at each temperature.

Fig. S4. Simulated marginal and joint distributions of phenotypic and genetic sex tendency under the threshold model of polygenic sex determination in A) a sex-neutral environment and B) a masculinizing, high temperature environment. Heritability (h^2) of phenotypic sex tendency was set to 0.60. When phenotypic sex tendency is positive, all animals differentiate as females, while they differentiate as males when phenotypic sex tendency is negative. In a sex-neutral environment, the mean phenotypic sex tendency of the population is zero, and thus the sex-ratio is 50:50. High temperature displaces the mean phenotypic sex tendency of the population to the left, resulting in lower proportion of females (here 30%). Note that genetic sex tendency has a lower variance than phenotypic sex tendency, as h²<1. Also note that h²<1 also implies that only animals with high genotypic sex tendency always differentiate as females, this constraint being even higher in the high temperature environment (B). A total of 5000 individuals were simulated, with environmental variance of sex tendency VE=1.0, additive genetic variance V_A=1.5 and phenotypic variance V_P=V_A+V_E=2.5, which corresponds to $h^2 = V_A/V_P = 0.60$.

Fig. S5. Circular Manhattan plots of genome wide association scan (GWAS) for sex (male/female) with, from the center to the border: the percentage of variance explained in a sliding window of 50 adjacent SNPs, the SNP effects in *-log(pvalue)* computed after 2 iteration of weighted GWAS, and the location of the genes previously reported as having a significant differential expression level in link with "sex differentiation", "sex determination" and "histone modification" in the present study. Red dots represent SNP markers explaining over 2% of the variance, and light blue shaded areas represent the regions including the putative quantative trait loci.

Biological Processes
Flexion

Fig. S6. Dotplot showing enrichment of Gene Ontology biological processes for up and downregulated DEGs between low (LT) and high temperature (HT) of whole fish at the "flexion" stage. Data are from the RNA-Seq. The size of the circle represents the number of genes enriched in the entry, and the color indicates the significance of the p-value.

Fig. S7. Dotplot showing enrichment of Gene Ontology biological processes for up and downregulated DEGs between females (positive correlation with eGST) and males (negative correlation with eGST) in whole fish at the "all fins" stage (53 and 78 dph, all temperature treatments cofounded). Data are from the RNA-Seq. The size of the circle represents the number of genes enriched in the entry, and the color indicates the significance of the p-value.

Fig. S8. Significant (p < 0.01) linear correlation between the estimated genetic sex tendency (eGST) and 3 genes (number of transcripts on the y axis) involved in the GO steroids metabolic process, namely *hsd17b1, cyp26a,* and *3β-hsd,* from the RNA-Seq analysis of the gonads of fish at the juvenile stage. Circles represented fish kept at high temperature (HT) = 21 °C; and triangles those kept at low temperature (LT) = 16 °C. Individuals are represented with a color gradient, from maroon to yellow, representing their eGST.

Fig. S9. Heatmap of all genes differentially expressed in gonads of fish at the juvenile stage between genetic males (eGST < 0) and genetic females (eGST > 0) in the gene ontology "sex determination". Individuals are represented with a color gradient, from maroon to yellow, representing their eGST. The phenotypic sex is assessed based on the automatic threshold proposed by the pheatmap package, with indivividuals represented in maroon considered as phenotypic males and those in yellow considered as phenotypic females. The fish from the high temperature (HT) tretament = 21 °C are represented in red; and those from the low temperature (LT) treatment = 16 °C are represented in blue.

Biological Processes

Fig. S10. Dotplot showing enrichment of Gene Ontology biological processes for up and downregulated DEGs between females (positive correlation with eGST) and males (negative correlation with eGST) in the gonads of fish at the juvenile stage (117 and 124 dph, all temperature treatments cofounded). Data are from the RNA-Seq. The size of the circle represents the number of genes enriched in the entry, and the color indicates the significance of the p-value.

Fig. S11. Significant (p < 0.01) linear correlation between the estimated genetic sex tendency (eGST) and both insr and sox9a (relative number of transcripts on the y axis). For five genes, *cyp19a1a, foxl2, dmrt1, amh, gsdf*, and the PC1 axis, a dichotomic distribution was observed and modelled with a "quasibinomial" function. Note that the correlation is still detectable with the outlier (extreme positive value at LT). Circles represented fish kept at high temperature (HT = 21 °C; n = 21); and triangles those kept at low temperature (LT = 16 °C; n = 22). Individuals are represented with a color gradient, from maroon to yellow, representing their lower or higher eGST.

Table S1. Genes from different GOs that are significantly correlated to the eGST in fish at the "all fin" stage. (the direction of the effect provide the direction of the correlation (negative = male and positive = females)

Sex differentiation Sex Determination Sex Determination Histone modifications **gene direction of the effect p-value gene direction of the effect p-value gene direction of the effect p-value7** bcl2 -631,78 0,0084 insr -263,0 0,0003 paxip1 44,34 0,003 insr -263,00 0,0003 six4 35,9 0,0018 prmt5 223,46 0,006 six4 35,88 0,0018 foxl2 246,0 0,0066 wac -128,71 0,008 foxl2 245,97 0,0066 sox9 -350,0 0,0067 mtf2 165,80 0,007 ptx3 -223,55 0,0037 dmrt1 -292,7 0,0065 pcgf6 21,08 0,008 tcf7 367,51 0,0066 tmem184a 32,5 0,0032 parg 145,11 0,006 tgfb2 -779,96 0,0092 pcgf5 - -38,01 0,003 eif2b4 55,26 0,0038 hdac1 208,51 0,002 rdh10 256,98 0,0023 zmpste24 173,91 0,000 sox9 -349,98 0,0067 asxl2 -115,91 0,006 dmrt1 -292,73 0,0065 prkd2 -85,95 0,002 amh -13130,47 0,0047 ppp5c 88,36 0,008 lgr4 -81,47 0,0096 ruvbl2 74,33 0,003 gas2 -330,17 0,0035 nsd1 -74,40 0,007 cbx2 105,19 0,0047 lmna 230,97 0,001 tbc1d20 61,47 0,0037 brms1l 29,99 0,003 angpt1 -438,16 0,0057 ash2l 86,93 0,001 ago4 -95,22 0,0055 setd5 -268,17 0,005 jmjd1c -221,58 0,0015 tada3 45,48 0,010 aromata se 1991,61 0,0045 hdac7 -166,63 0,002 auts2 -260,43 0,002 kansl1 -224.33 0.010 fmr1 82,03 0,006 hopx -191,18 0,009 pcgf1 42,06 0,010 ccna2 616,58 0,004 ctbp1 -223,82 0,001 smarcad1 54,69 0,002 noc2l 92,77 0,009 suv39h1 25,43 0,002 eed 63,74 0,000 men1 57,21 0,008 uhrf1 339,56 0,002 actl6a 140,25 0,001 mier2 34,26 0,009 bcl6 -144,20 0,004 gfi1 -20,71 0,008

Table S2. Genes from different GOs that are significantly correlated to the eGST in the gonad of juveniles fish. (the direction of the effect provide the direction of the correlation (negative = male and positive = females)

Table S3. Descriptive information regarding sampling

juveniles stage - gonades

Legends for Datasets S1 to S4

Dataset S1: DESeq2 Normalized counts of fish at the flexion stage

Dataset S2: DESeq2 comparison between 16 and 21 genes differentially expressed for fish at the flexion stage

Dataset S3: DESeq2 Normalized counts of fish at the all fin stage

Dataset S4: DESeq2 Normalized counts of fish at the early juvenile stage

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Sample References:

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