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Liver proteomics of gilthead sea bream (Sparus aurata) exposed to cold stress

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17 Abstract

18 The gilthead sea bream (Sparus aurata, L.) is very sensitive to low temperatures, which induce fasting and 19 reduced growth performances. There is a strong interest in understanding the impact of cold on fish 20 metabolism to foster the development and optimization of specific aquaculture practices for the winter period. In this study, a 8 week feeding trial was carried out on gilthead sea bream juveniles reared in a 21 22 Recirculated Aquaculture System (RAS) by applying a temperature ramp in two phases of four weeks each: a cooling phase from 18°C to 11°C and a cold maintenance phase at 11°C. Liver protein profiles were 23 24 evaluated with a shotgun proteomics workflow based on filter-aided sample preparation (FASP) and liquid chromatography-mass spectrometry (LC-ESI-O-TOF MS/MS) followed by label-free differential analysis. 25 Along the whole trial, sea breams underwent several changes in liver protein abundance. These occurred 26 mostly during the cooling phase when catabolic processes were mainly observed, including protein and lipid 27 degradation, together with a reduction in protein synthesis and amino acid metabolism. A decrease in protein 28 mediators of oxidative stress protection was also seen. Liver protein profiles changed less during cold 29 maintenance, but pathways such as the methionine cycle and sugar metabolism were significantly affected. 30 31 These results provide novel insights on the dynamics and extent of the metabolic shift occurring in sea bream 32 liver with decreasing water temperature, supporting future studies on temperature-adapted feed formulations. The mass spectrometry proteomics data have been deposited to the ProteomeXchange Consortium via the 33 PRIDE partner repository with the dataset identifier PXD011059. 34

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36 Keywords: cold stress; winter syndrome; gilthead sea bream; liver proteins; methionine; shotgun

- 37 proteomics.
- 38

39 **1. Introduction**

40 Gilthead sea breams (Sparus aurata, L.) have their natural habitat in the Mediterranean Sea at water temperatures ranging seasonally from approximately 26°C in summer to 11-13°C or even lower in winter, 41 42 depending on the specific environment. Wild sea breams manage this temperature decrease by migrating to deeper and warmer water (Davis, 1988). In fact, water temperatures below 13°C lead to both behavioral 43 (e.g., erratic swimming, voluntary fasting, hyposensitivity to stimuli) and physiological (e.g., impaired 44 growth, fatty liver, tissue necrosis, infections) stressful changes that can ultimately lead to physiological 45 46 dysfunction and death (Contessi et al., 2006; Gallardo et al., 2003; Ibarz et al., 2010b). However, farmed sea breams living in outdoor tanks and in floating sea cages are unable to avoid this thermal stress, and their 47 prolonged exposure to temperatures below 13°C causes decrease in activity (Ibarz et al., 2003), growth delay 48 (Tort et al., 1998), metabolic depression (Ibarz et al., 2018; Sanahuja et al., 2019) and reduced feed 49 consumption until total fasting when water temperatures fall below 10°C (Ibarz et al., 2010b). Other 50 physiological alterations include hepatic functionality, with the liver becoming steatosic and whitish due to a 51 large deposition of lipids, as well as reduced efficiency of adaptive immunity with increased susceptibility to 52 53 infections (winter syndrome or winter disease) and alteration of the main redox pathways (Abram et al., 2017; Ibarz et al., 2007, 2005; Sánchez-Nuño et al., 2018). These phenomena impact farming productions 54 causing relevant economic losses with a consequent strong interest of fish farmers in finding efficient 55 56 strategies for their reduction. One way for compensating thermal stress is represented by enhancing the nutritional state and metabolism through the use of feeds specifically designed for the colder season, and this 57 58 requires understanding the consequences of cold on fish metabolism. To this aim, proteomic, metabolomic 59 and transcriptomic approaches have been applied to farmed and wild fish biofluids and tissues, such as serum, liver, muscle and other organs, with differing degrees of success (Addis, 2013; Addis et al., 2010a, 60 2010b; Alves et al., 2010; Braceland et al., 2013; Brunt et al., 2008; Douxfils et al., 2011; Ghisaura et al., 61 62 2014; Martin et al., 2001; Melis et al., 2017; Mininni et al., 2014; Rodrigues et al., 2012). Proteomics offers several specific advantages when compared to genomic or transcriptomic strategies, including the ability to 63 evaluate the actual extent of protein abundance, going beyond the estimates based on gene expression. In 64 fact, proteins do not always follow a strict relationship with gene transcription but are regulated at the 65 66 translational or post-translational level. Liver, the main metabolic organ of the body, has gained the greatest

67 attention in proteomic studies evaluating the influence of farming practices on fish metabolism. Martin and coworkers (Martin et al., 2001) studied the changes occurring in the liver proteome as a consequence of 68 69 different feeding regimens, including dietary plant protein substitution. Liver metabolism is considerably 70 influenced also by other factors including environmental stress, and might be affected by xenobiotics and toxins (Addis, 2013; Ghisaura et al., 2014). Numerous research groups focused on the gilthead sea bream 71 liver proteome to investigate a variety of stressful factors, ranging from handling and crowding (Alves et al., 72 73 2010) to the use of antiparasitic, or antibacterial agents and different environmental pollution (Isani et al., 2011; Kovacik et al., 2018; Varó et al., 2013). Proteomics has also been used to assess the impact of cold on 74 75 fish metabolism (Ibarz et al., 2010a; Parrington and Coward, 2002; Vilhelmsson et al., 2003). However, these studies applied a gel-based approach (2D-gel electrophoresis, 1D GelC-MS/MS). With the aim of 76 77 gathering additional information, we investigated the changes occurring in liver tissue by applying a shotgun proteomics workflow based on filter-aided sample preparation (FASP), tandem mass spectrometry (MS/MS), 78 79 label-free quantitation and, finally, pathway analysis by means of STRING and Ingenuity Pathway Analysis (IPA). In fact, this approach provides a higher proteome coverage and is less affected by the typical 80 81 limitations of gel-based studies, especially when 2D separation is involved (Westermeier et al., 2008), and additional or complementary information can be obtained in respect to previous studies. Liver proteome 82 changes were assessed by mimicking the winter challenge conditions in a Recirculated Aquaculture System 83 84 (RAS), both immediately after temperature reduction and during cold maintenance.

85

86 2. Materials and Methods

87 2.1. Experimental design

A detailed description of the trial can be found in our previous work (Melis et al., 2017). For the purposes of both studies, a total of 60 juvenile gilthead sea breams with an average weight of 82.0 ± 4.5 g were uniformly distributed in three fiberglass tanks of 550 L with mechanical and biological filtration systems, a pumping system, a water thermoregulation system, and an automatic control for adjusting and monitoring the main physicochemical parameters. Fish were first acclimated for two weeks by linearly lowering water temperature from 20°C to 18°C (t0). Then, temperature was reduced at a rate of approximately 1°C every Tuesday and Friday until reaching 11°C (cooling phase, t0-t1, 4 weeks). Then, fish were maintained at 11°C

95 (cold maintenance phase, t1-t2) for the same time span (4 weeks). Fish were fed by hand, once a day, an experimental feed formulation (Aller Aqua, Christiansfeld, Denmark) with 43% protein and 14% fat. Feed 96 97 ration was adjusted accounting for fish size, biomass and temperature (0.9 ± 0.03 g during cooling and $0.4 \pm$ 98 0.03 g during cold maintenance). During the growth trial no mortality occurred, and fish never stopped eating below 13°C. For the purposes of this study, at the beginning of the trial (t0) and at each time point (t1, 99 t2), 9 fishes were anesthetized with 1,1,1-trichloro-2-methylpropan-2-ol (2% in marine water) and 100 101 transferred in a mixture of marine water and ice (total number of sacrificed subjects = 27). Liver was excised from each fish, weighed and frozen rapidly in liquid nitrogen in Petri dishes as described by Melis and 102 coworkers (2017). 103

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105 **2.2. Protein extraction and quantification**

Fish liver protein extraction was performed according to Ghisaura et al. (2016). Briefly, a small portion of 106 each tissue (100 mg) was placed in a 2 ml Eppendorf safe-lock tube (Eppendorf, Hamburg, Germany) and 107 immersed at 25% w/v in lysis buffer (7 M urea, 2 M Thiourea, 2% CHAPS) plus protease inhibitor cocktail 108 109 (Protease Inhibitor Cocktail for General Use, Sigma-Aldrich, Saint Louis, MO) as indicated in the manufacturer instructions. Samples were then processed and subjected to three cycles of 5 min at 30 110 oscillations/s in a TissueLyser mechanical homogenizer (Qiagen, Hilden, Germany). Samples were frozen in 111 112 between homogenization cycles to ease tissue disruption and avoid excessive sample heating. Protein 113 extracts were then centrifuged for 15 min at 18,000 x g at 4°C, quantified with the Pierce 660 nm Protein Assay Kit (Thermo Scientific - Rockford, IL), evaluated for quality and integrity by SDS-PAGE (data not 114 shown), and stored at -80°C until use. 115

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117 2.3. Shotgun proteomics

Three protein samples for each time point (t0, t1, t2), each constituted by a biological pool of three fish liver extracts, were used for shotgun proteomics analysis. Protein extracts (n = 9) were subjected to onfilter reduction, alkylation, and trypsin digestion according to the filter-aided sample preparation (FASP) protocol (Wiśniewski et al., 2009), with minor modifications (Ghisaura et al., 2016; Tanca et al., 2013) using Amicon Ultra-0.5 centrifugal filter units with Ultracel-10 membrane (Millipore, Billerica, MA,

123 USA). Peptide mixture concentration was estimated by using BCA protein assay kit (Thermo Scientific -Rockford, IL). LC-MS/MS analyses were performed on a Q-TOF hybrid mass spectrometer with a nano 124 125 lock Z spray source, coupled on-line with a NanoAcquity chromatography system (Waters) (Pagnozzi et al., 2014). Two technical replicates were analyzed for each biological pool (total LC-MS/MS runs = 18). 126 Peptide mixtures were concentrated and washed with an enrichment column and then fractionated over a 127 250 min gradient on a C18 reverse phase column. The instrument was set up in a data-dependent MS/MS 128 129 mode, with a full-scan spectrum followed by tandem mass spectra, selecting peptide ions as the three most intense peaks of the previous scan. ProteinLynx software (Version 2.2.5), was used to produce the peak 130 lists as pkl files. 131

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133 2.4. Protein identification and differential proteomic analysis by label-free quantitation

The Q-TOF peak lists were analyzed by Proteome Discoverer software (version 1.4; Thermo Scientific), 134 after conversion into MGF files. Technical replicates were processed as merged, generating one list of 135 identified proteins for each biological sample by the Proteome Discoverer Daemon utility. The workflow 136 137 was made up of the following nodes (and respective parameters). Spectrum Selector for spectra preprocessing (precursor mass range: 350-5000 Da; S/N Threshold: 1.5), Sequest-HT as search engine 138 (Protein Database: Chordata sequences from UniProtKB; Enzyme: Trypsin; Max. missed cleavage sites: 2; 139 140 Peptide length range 5–50 amino acids; Max. Delta Cn: 0.05; Precursor mass tolerance: 50 ppm; Fragment 141 mass tolerance 0.4 Da; Static modification: cysteine carbamidomethylation; Dynamic modification: methionine oxidation), and Percolator for peptide validation (FDR < 1% based on peptide q-value) (Choi 142 and Nesvizhskii, 2008; Käll et al., 2009, 2008; Spivak et al., 2009). In order to estimate the extent of 143 differential protein abundance among sample groups, the Normalized Spectral Abundance Factor (NSAF) 144 145 was calculated for each protein according to Zybailov et al. (Zybailov et al., 2006) as follows:

146 NSAF=SAFi/ $\sum_{i=1}^{N} SAFi$, where subscript *i* indicates a protein identity, N represents the total number of 147 proteins, and SAF is a protein spectral abundance factor (protein spectral counts divided by its length). 148 Finally, the NSAF log ratio (R_{NSAF}) was calculated as follows: R_{NSAF} = log₂(NSAF_x + CF)/(NSAF_y + CF), 149 where NSAF_x and NSAF_y are the summed NSAF values for each protein in sample groups to be compared 150 (x = t1 or t2; y = t0 or t1, respectively) and CF is a correction factor, empirically set to 2 (Tanca et al.,

151 2015, 2012). Statistical significance of differential protein abundance was further assessed by applying the 152 Student's *t*-test (two-sample comparison, p < 0.05) on logarithmic NSAF values, after replacing missing 153 values with 0.1 (empirically determined as in Zybailov et al., 2006) and corrected by using false discovery 154 rate (FDR) as a multiple hypothesis testing, with FDR < 0.1 as a threshold limit. Only proteins with $R_{NSAF} >$ 155 0.5 or < -0.5 were considered. The mass spectrometry proteomics data have been deposited to the 156 ProteomeXchange Consortium via the PRIDE partner repository with the dataset identifier PXD011059 157 (Deutsch et al., 2017; Sanchez et al., 2015; Vizcaíno et al., 2016) and are reported in Ghisaura et al., 2019.

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159 **2.5. Multivariate data statistical analysis**

Multivariate statistical data analyses (MVDA) were made using SIMCA-P 13.0 version (Umetrics, Inc., 160 Kinnelon, NJ). Prior to analysis, NSAF values were subjected to log transformation and Pareto scaling. For 161 multivariate preliminary inspection, an unsupervised principal component analysis (PCA) was performed, 162 followed by supervised OPLS-DA (orthogonal partial-least square discriminant analysis) to display as score 163 plot the NSAF clustering, according to each temperature variation. Goodness of all MVA models was 164 165 evaluated by the cumulative R2(cum) and the predictive R2Y(cum) and Q2(cum) parameters, calculated according to the cross-validation method. In particular, R2Y(cum) is defined as the proportion of variance in 166 the data explained by the models and indicates the goodness of fit, whereas Q2(cum) is defined as the 167 proportion of variance in the data predictable by the model. Both R2Y(cum) and Q2(cum) vary between 0 168 and 1: a good prediction model is indicated by Q2(cum) > 0.5, whereas a Q2(cum) > 0.8-0.9 means an 169 excellent predictive ability of the model; for Q2(cum) values close to 0.5, no statistical group separation 170 between observed clusters was considered (Westerhuis et al., 2008). 171

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173 **2.6. Pathway analysis**

Gene ontology and protein annotations were retrieved from UniProtKB (http://www.uniprot.org). The 174 175 uncharacterized sequences blasted **NCBI** non-redundant were on database: 176 (http://blast.ncbi.nlm.nih.gov/Blast.cgi) to find the homologous proteins. For pathway analysis two online software packages were used: STRING version 10 (Search Tool for the Retrieval of Interacting 177 Genes/Proteins; http://string-db.org) (Szklarczyk et al., 2017) and the online software package IPA (version 178

179 9.0; Ingenuity Systems, Redwood City, CA). For STRING, an enrichment analysis was performed with all differentially expressed proteins, and KEGG pathways, GO Biological Processes and GO Molecular 180 181 Functions implemented in the web platform were investigated by using the *Danio rerio* as organism model. Only the pathways and molecular networks displaying an FDR < 0.05 were considered as significantly 182 enriched in the protein list and were considered for further analyses. For IPA analysis, since the software 183 operates on a database built on the literature generated for humans and rodents, fish UniProt IDs were 184 185 replaced with the closest mouse (Mus musculus) protein equivalents to enable a wider knowledge-based investigation of pathways as previously described (Addis et al., 2011; Ghisaura et al., 2014; Terova et al., 186 2014), being this a larger and better investigated database. The list of protein identifications (IDs), with 187 their respective R_{NSAF} and p values (< 0.05) were used for the gene ontology analysis. Diseases and 188 functions were specifically considered to focus on the physiological and health state of the fish liver during 189 cold treatment. 190

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192 **3. Results and Discussion**

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194 **3.1.** Cooling phase

A preliminary evaluation of experimental NSAF data with PCA indicated that liver proteomes at t1 were clearly separated from those at t0 (**Figure S1**). This result was confirmed by a further supervised approach based on OPLS-DA (**Figure 1a**). Indeed, the OPLS-DA model showed a very good fit parameter (R2Y(cum) = 0.819) and a fairly good prediction ability (Q2(cum) = 0.828), as shown by the model validation results (**Figure 1**).

A total of 42 proteins showed statistically significant differences in abundance at t1 *vs* t0 ($R_{NSAF} > 0.5$ or < -0.5, p < 0.05, FDR < 0.1) and are listed in **Table 1**. According to STRING, numerous metabolic pathways were affected, as reported in **Table S1**. KEGG pathways were mainly related to carbon metabolism such as amino acid metabolism, including the phenylalanine, tyrosine and cysteine and methionine metabolisms. Other general pathways that encompass metabolic pathways and carbon metabolism were also statistically significant (FDR <0.05). Amino acid metabolism plays an important role in fish metabolism for protein synthesis, glucose formation, and energy. For example, Costas et al, (2011) observed increased levels of

207 amino acids and other metabolites during long term feed deprivation in Senegalese sole fish, and amino acids are considered the major source of energy in this carnivorous fish species. Feed deprivation led to active 208 209 gluconeogenic active processes in the liver supported by proteolysis in 21 days feed-deprived sole, 210 suggesting that amino acids are employed as a carbon source for gluconeogenesis for the maintenance of plasma glucose levels. Phenylalanine and tyrosine can influence pigmentation, development, feed intake, 211 growth performance, immunity, and survival of fish in the natural environment. Tyrosine synthesized from 212 213 the essential amino acid phenylalanine is a precursor for important hormones and neurotransmitters, including thyroid hormones that play an important role next to energy metabolism and protein synthesis 214 215 (Jasour et al., 2017; Li et al., 2009).

According to IPA, increased proteins were mainly associated to cellular stress and to protein and lipid 216 217 degradation processes, while decreased proteins were mostly related to protein synthesis, actin-binding activity, amino acid metabolism, and protection from oxidative stress (Table S2). Several enzymes included 218 in amino acid metabolism showed significant changes. An increase in fumarylacetoacetase and a reduction in 219 4-hydroxyphenylpyruvate dioxygenase were observed (Table 1), both enzymes being involved in the 220 221 catabolism of phenylalanine and tyrosine. Downregulation of phenylalanine and tyrosine catabolism has been associated to liver damage (Richard et al., 2016), confirming the key role of these amino acids in the 222 physiological response of sea bream to cold challenge. Catabolism of specific amino acids, including 223 224 tyrosine and phenylalanine through homogentisate 1,2-dioxygenase, was also observed to be reduced under 225 cold stress by Ibarz and coworkers (2010a). This has been associated to the possible entrance of tyrosine and phenylalanine in the TCA cycle to produce energy. Likely, it has been suggested that high dietary 226 227 availability of amino acids, methyl donors (betaine, choline) and cofactors (folate) supports the flux toward the methionine cycle thus favoring optimal homeostasis, helping fish to cope with exogenous stress and 228 229 finally improving feed performance (Richard et al., 2016). Amino acids can be considered an important 230 glucogenic source in fish. In this regard, high inclusion levels of feathermeal in feed have been correlated with hepatic levels of leucine, isoleucine, tyrosine, valine, methionine, arginine, and phenylalanine involved 231 232 in energy metabolism. Moreover, an increased concentration of these amino acids in the liver was observed 233 in the same study, indicating their inhibition from entering the TCA metabolic pathway to generate energy 234 (Jasour et al., 2017). Formimidoyltransferase-cyclodeaminase was also increased. This enzyme is involved in

235 the sub-pathway that synthesizes glutamate by means of histidine degradation and tetrahydrofolate conversion in the pathway of one carbon metabolism. Thioredoxin and metallothionein were reduced. These 236 237 proteins act as primary liver defense under oxidant attack at low temperatures. Their reduction might be 238 associated to loss of hepatic functionality under cold stress (liver failure), confirming that oxidative stress and amino acid metabolisms are the pathways mainly affected by cold. Previous results by Ibarz et al. 239 (2010a) are in line with our findings. In their proteomic study on gilthead sea bream liver under acute cold 240 241 challenge, they identified oxidative stress, amino acid metabolisms and carbohydrate metabolism as the most 242 perturbed pathways.

Actin-related proteins such as cofilin-2 were reduced during water cooling. Similarly, Ibarz and coworkers 243 (2010a) found decreased actin levels in cold-stressed fish. Both alpha and beta tubulins, the components of 244 245 cytoskeletal microtubules, were increased, as already observed in previous studies on cold challenged gilthead sea breams. The same authors suggested that tubulin has a protective effect against cold stress, but 246 the exact mechanisms are still under study. The evident increase seen in proteasome-associated proteins, 247 together with the observed changes in cytoskeletal proteins, might suggest the occurrence of tissue 248 249 remodeling processes induced by thermal stress. A reduction in heart-like FABP or FABP3 was observed (Table 1). Mininni and coworkers (2014) reported several changes in FABP isoforms according to 250 251 transcriptomics, but did not report changes in FABP3.

Several "diseases and functions" categories were also significant according to IPA (**Table S2**). These results, although obtained by comparison of literature based on mouse, support the finding that the most relevant changes in liver metabolism occurred during the cooling phase. As summarized in **Table S2**, it is worth noting that several significantly modified proteins were involved in cellular growth and proliferation, inflammatory response, and infectious diseases. This may suggest some similarity with the well-known effects of winter syndrome in gilthead sea bream (Ibarz et al., 2010b).

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259 **3.2. Cold maintenance phase**

In the cold maintenance phase (constant 11°C, comparison t2 *vs* t1), liver proteins showed less marked changes than in the cooling phase. Consequently, the t2 and t1 sample clusters displayed a lower PCA separation (**Figure S1b**). Quantitative estimation of discriminative values of the model is given by lower

263 Q2(cum) index (Q2(cum) = 0.55), close to the threshold limit for a biological model discrimination 264 (Westerhuis et al., 2008). Similarly, the goodness of fit of the related OPLS-DA model led to a lower 265 Q2(cum) value (Q2 (cum)_{COLDMAINT} = 0.69, **Figure 2**) when compared to the cooling phase (Q2(cum)_{COOLING} 266 = 0.83, **Figure 1**). These observations indicate a lower prediction ability of all the MVDA models associated 267 to the cold maintenance phase *vs* the cooling phase, and suggest that t2 and t1 (that is, along cold 268 maintenance) are more similar to each other (**Figure S1b**) than t0 and t1 (beginning and end of cooling 269 phase).

In this case, only 24 significantly differential proteins were identified in all samples (R_{NSAF} >0.5 or <-0.5, p < 0.05, FDR < 0.1) and are listed in **Table 2**. Several proteins were associated to the methionine cycle, such as betaine-homocysteine-S-methyltransferase (BHMT). Interestingly, this protein underwent the most intense change in abundance observed in the whole study (R_{NSAF} = 2.96). Other proteins with ribosomal activity, glycolytic and gluconeogenetic function, were increased in t2 vs t1, whereas proteins associated to scavenger activity, migration and cellular organization, protein transport and proteolysis, adenosine and homocysteine synthesis were decreased (**Table S1**).

Enrichment analysis indicated the main KEGG pathways altered during the cold maintenance phase (**Table S1**). Among them, the categories metabolic pathways and glycolysis/gluconeogenesis had three molecules involved. Amino acid metabolisms, including alanine, aspartate and glutamate metabolism, cysteine and methionine metabolisms, were also significant, confirming the key role of energy metabolism and of the methionine cycle. KEGG pathways, GO Biological Processes, and GO Molecular functions are fully detailed in **Table S1**. Only two diseases and function features were associated to the cold maintenance phase (**Table S2**), supporting the above observations.

In a previous investigation on the metabolic response of sea breams to low water temperatures (Melis et al., 2017), liver gluconeogenesis was more affected than glycolysis during constant cold temperatures. In this study, an increased abundance of proteins implicated in carbohydrate metabolism was observed in the same conditions. Also, the observed perturbation of amino acid metabolism during the cold maintenance phase is confirmed by previous proteomic investigations (Richard et al., 2016). In particular, the methionine cycle and several related molecules (betaine, choline/phosphocholine and glutathione) were found to be affected during prolonged exposure to low temperatures. Several scientific reports on cold challenged gilthead sea

291 bream highlight a methionine cycle activation under stress conditions (Ibarz et al., 2010; Mininni et al., 292 2014; Richard et al., 2016). Amino acid interconversion and catabolism processes, often resulting in 293 upregulation of BHMT, are likely triggered during cold stress to prevent hepatic accumulation of cytotoxic 294 molecules such as homocysteine. However, although fast water temperature decrease such as 20°C to 8°C in 3 days (Ibarz et al., 2010a) can lead to downregulation of BHMT, slower cooling conditions seem to 295 296 consistently result in an increased expression. Our previous work (Melis et al., 2017), based on a first gradual 297 temperature decrease followed by a cold maintenance phase, reached a similar conclusion, i.e. adjustments of methionine cycle metabolism are activated during prolonged stress while they are not observed during the 298 299 first, more acute phases of temperature decrease. In this sense, BHMT appears as a key proteomic regulator throughout a persistent stressful condition, acting via antioxidant mechanisms by balancing S-adenosyl-300 301 methionine and preventing toxic homocysteine accumulation. The increase in BHMT was also associated to an unbalanced amino acid composition of diets leading to increased oxidative stress (Ghisaura et al., 2014), 302 which suggests that BHMT is a good indicator related to general fish homeostasis. The changes in BHMT 303 levels might therefore be related also to glutathione biosynthesis, reflecting a different extent of oxidative 304 305 stress caused on hepatocytes by the different temperatures; more specifically, its increase in the cold maintenance phase (t2 vs t1) indicates that a higher oxidative stress is exerted by long exposure to winter 306 temperatures (Ghisaura et al., 2014). Other amino acid degradation pathways and changes in proteins mainly 307 implicated in stress response and cellular defense were represented by Adenosylhomocysteinase (AHCY or 308 309 SAM) that is involved in the superpathway of methionine degradation, as well as in the methionine salvage pathway; specifically, it catalyzes the reversible hydrolysis of S-adenosylhomocysteine to adenosine and L-310 homocysteine and has a central role in the regulation of methyltransferase reactions, important for liver 311 homeostasis maintenance (Ghisaura et al., 2014). As pointed out by Richard et al., (2016) its deficiency is 312 313 usually associated with hepatic damages.

Increased purine metabolism also plays a crucial role in facing cold stress. Its involvement was observed in both muscle and liver of gilthead sea bream during cold maintenance phases by metabolomics (Melis et al., 2017) and was confirmed here by the observed increase in purine nucleoside phosphorylase-like (PNP). It was also postulated that the involvement of purine metabolism pathways in cold challenged gilthead sea

bream might be somewhat associated to other symptoms of the winter syndrome, such as skin pigmentationand immune suppression (Melis et al., 2017).

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321 **3.3.** Overall changes

Proteomic profiles at the starting time (t0) were then compared with those at the end of the trial (t2) to gain a general overview of the changes occurring along prolonged thermal stress. A very clear separation of the two conditions was highlighted by the PCA score plot (**Figure S1c**) showing a Q2(cum) = 0.68. Furthermore, OPLS-DA score plot and related model validation (**Figure 3**) showed an excellent cluster separation with a Q2(cum) closer to 0.9 (Q2(cum) = 0.885), which implies a higher goodness of discrimination with respect to both t1/t0 (**Figure 1**) and t2/t1 (**Figure 2**).

328 In this case, 59 proteins showed differences in abundance ($R_{NSAF} > 0.5$ or < -0.5, p < 0.05, FDR < 0.1) and 329 are listed in Table 3. Proteins related to proteolysis processes, energy conversion, carbohydrate and amino acid metabolism and mitochondrial activity were increased, while those involved in fatty acid metabolism 330 and amino acid conversion were decreased, as well as proteins associated to protection from oxidative stress 331 332 and purine degradation (Table S1). Several stress-driven metabolic changes were observed. Abundance variations affected proteins involved in the metabolism of key amino acids (histidine, alanine, aspartate and 333 glutamate; tyrosine, phenylalanine, arginine and proline; cysteine and methionine). Full KEGG pathways, 334 335 GO Biological Processes and GO Molecular functions are detailed in Table S1. The categories pyruvate 336 metabolism and starch and sucrose metabolisms were also involved. Moreover, five significantly increased diseases and functions categories were represented by inflammatory responses of the organism, generation of 337 reactive oxygen species (and related free radical scavenging functions), liver necrosis, lipid metabolisms and 338 their oxidation (functional to energy production) (Table S2). 339

The analysis of the overall cold challenge confirmed the observations done on the two separate phases (cooling and cold maintenance). Upon constant temperature decrease, proteins implicated in proteolysis, lipolysis, glycolysis and glycogenolysis, together with those implicated in amino acid metabolism, increased in abundance, possibly to intensify energy production during cold stress. In fact, as seen in many studies (Chang et al., 2018; Ibarz et al., 2010b, 2010a; Richard et al., 2016), there is a strong mobilization of extrahepatic fat deposits to liver and glycogen reserves. Amino acid degradation pathways (phenylalanine

346 and tyrosine catabolism) and changes in proteins mainly implicated in stress response and cellular defense were represented by a slightly lowered uricase (Table 3), that catalyzes the oxidation of uric acid to 5-347 348 hydroxyisourate and then to allantoin, a degradation product of purine nucleobases. Table S2 reports diseases and functions categories affected by cold stress exposure. Cold mainly affected inflammatory and 349 organ damage processes, oxidative stress response and tyrosine degradation. The involvement of this latter 350 pathway was recently confirmed by our research group using ¹H NMR-based metabolic fingerprinting (Melis 351 352 et al., 2017) and by Richard and coworkers (Richard et al., 2016) who specifically described phenylalanine and tyrosine catabolism and their interconversion in the proteomic response of gilthead sea bream to low 353 temperature. 354

355

4. Conclusions

The protein makeup of sea bream liver undergoes several changes upon exposure to decreasing water 357 358 temperature, suggesting the occurrence of a metabolic shift enabling adaptation to changed environmental 359 conditions. This shift occurs mainly along temperature lowering in a cold adaptation phase. The maintenance 360 of low but constant temperatures seems to affect protein levels to a lesser extent, although significant changes emerge also in this phase, such as in methionine metabolism. Gaining a greater knowledge of sea 361 bream metabolic changes to cold adaptation might be of use to fish farmers for the development of specific 362 aquaculture practices aimed at mitigating the negative effects of cold on fish growth, including the design of 363 novel feed formulations for the winter season. 364

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366 Acknowledgements

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Table 1

Sea bream liver proteins undergoing significant changes during the cooling phase (t1 *vs* t0). $R_{NSAF} > 0.5$ or < -0.5; p value < 0.05; FDR multiple comparison test <0.1.

Accession number	Protein name	R _{NSAF} t1/t0
Increased protein	S	
O4RBW9	Proteasome subunit beta type-2	2.0501
B3F9U6	Hemoglobin beta chain	1.7291
O1PCB2	Beta globin	1.6826
P86232	Ezrin (Fragments)	1.5108
P11748	Hemoglobin subunit alpha	1.3526
K7GAK5	Tubulin beta-7 chain	0.9969
04\$3.13	GTP-binding nuclear protein Ran	0.8656
091060	Tubulin alpha chain	0.8281
M9P052	Lysosomal acid lipase	0.8276
O4RVS0	ATP synthase F(0) complex subunit B1, mitochondrial	0.7537
L5M3T4	GTP-binding protein SAR1a	0.707
04\$798	Nucleolin isoform X2 (Fragment)	0.6757
H2MYW8	Fumarylacetoacetase	0.6599
J7FII7	Glutathione S-transferase (Fragment)	0.6588
G9I0G6	Transferrin	0.6428
S4S3W7	Phosphoglucomutase 1 (Fragment)	0.5869
I3JSE9	Formimidovltransferase-cyclodeaminase-like	0.5706
G10D60	H3 histone (Fragment)	0.5705
H2LS09	Nucleolin isoform X1	0.5106
D		
Decreased protein		o - - -
QUGPQ8	Cytochrome P450 2P11	-0.5457
A0A060VGE8	Cytochrome oxidase subunit II	-0.5506
W5LDH9	Uricase	-0.584
G3PTX7	Endoplasmic reticulum resident protein 27	-0.5926
HOYZDO	Electron transfer flavoprotein subunit alpha, mitochondrial	-0.6356
W5N925	Protein disulfide-isomerase (Fragment)	-0.6434
H2RKV3	Malic enzyme	-0.6595
M4AX90	Peroxisomal 2,4-dienoyl-CoA reductase-like	-0.6643
Q27HS3	Vascular smooth muscle alpha-actin (Fragment)	-0.6784
Q4RKE4	Fatty acid-binding protein, heart-like	-0.7223
Q8JHC5	Metallothionein (Fragment)	-0.7259
FIQ6EI	4-hydroxyphenylpyruvate dioxygenase	-0.7963
A0A060WA9	Adenosylhomocysteinase B	-0.8248
M4VQF0	Glyceraldehyde-3-phosphate dehydrogenase	-0.8662
M4AAN9	Phosphate carrier protein, mitochondrial-like isoform X1	-0.8739
F/DQ24	11-cis retinol dehydrogenase-like	-0.92
F/FYK5	40S ribosomal protein SA-like	-1.0889
AUAU60YQH0	Aspartate aminotransferase, cytoplasmic-like	-1.1398
B5X8Y0	Cofilin-2	-1.2505
H2VEH5	Peptidyl-prolyl cis-trans isomerase	-1.2816
B9EN58	Thioredoxin	-1.841
G3HK42	60S ribosomal protein L30	-2.441
G3UYV7	40S ribosomal protein S28 (Fragment)	-2.7022

379 **Table 2**

380

Sea bream liver proteins undergoing significant changes during the maintenance phase (t2 *vs* t1). R_{NSAF} > 0.5 or < -0.5; p value < 0.05; FDR multiple comparison test <0.1.

383

Accession number	Protein name	$R_{\rm NSAF} t2/t1$
. .		
Increased prote	eins	2.0525
I3KAPI	Betaine-homocysteine S-methyltransferase 1-like	2.9636
E9QBF0	Triosephosphate isomerase	1.4764
M/BNB0	60S ribosomal protein L30	1.4457
P61155	40S ribosomal protein S19	1.4016
F6Q602	Probable imidazolonepropionase	1.3571
C1KBH6	Phosphoenolpyruvate carboxykinase	1.3021
H2L7M4	Keratin, type I cytoskeletal 18-like	1.1887
F1QXV8	Phosphoglycerate kinase	1.0446
W5LXZ1	Purine nucleoside phosphorylase-like (Fragment)	0.9842
H2MFC0	Obg-like ATPase 1	0.9796
M4ANE8	Glutamate dehydrogenase, mitochondrial-like	0.766
G3Q9K3	ATP synthase subunit gamma	0.7505
C3KIP4	Myosin light polypeptide	0.7423
B5X124	Deoxyribose-phosphate aldolase	0.6882
I3KYC9	Fumarate hydratase, mitochondrial-like	0.6618
F1R0A9	Glucose-6-phosphate translocase isoform X1	0.5549
Decreased prot	teins	
M9NZ74	94 kDa glucose-regulated protein	-0.5354
L5M3T4	GTP-binding protein SAR1a	-0.6744
Q4QY80	Elastase 4-like protein (Fragment)	-0.9622
H2UYH6	60S ribosomal protein L6-like	-1.0566
H3CCF6	Delta-1-pyrroline-5-carboxylate dehydrogenase, mitochondrial	-1.1152
G5DYL0	Putative s-adenosylhomocysteine hydrolase (Fragment)	-1.294
H7C3T4	Peroxiredoxin-4 (Fragment)	-1.366
P86232	Fzrin	-1 4512

384 385 C

Table 3

Sea bream liver proteins undergoing significant changes along the whole trial (t2 *vs* t0). $R_{NSAF} > 0.5$ or < -0.5; p value < 0.05; FDR multiple comparison test <0.1.

Accession number	Protein name	$R_{\rm NSAF}$ t2/t
Increased protein	25	~
S7MY91	60S ribosomal protein L 12	2 434
O4RBW9	Proteasome subunit beta type-?	2.454
$O1PCB^{2}$	Beta globin	1 8357
P56251	Hemoglobin subunit beta	1.6591
P11748	Hemoglobin subunit alpha	1.5843
F90H32	Nucleoside diphosphate kinase	1.5045
M/ARN8	Mitochondrial pyruvate carrier 2-like	1.770
$F_{2VI} \Delta 1$	Transferrin	1.405
H2V638	Profilin	1.1002
	Mitochondrial 2 oxodicarboxylate carrier	1.0773
HOVMDO	Histone H2B	1.0445
K7GAK5	Tubulin heta-7 chain	0.993/
FOORFO	Triosenhosnhate isomerase	0.9793
$M/\Lambda INO$	L 2 hydroxyglutarate dehydrogenase, mitochondrial like	0.9793
RSYDR?	Inorganic pyrophosphatase 2 mitochondrial	0.9783
DJADKZ U2MWNIQ	Formimidoultransforase evaludouminase	0.7623
H21 S00	nucleolin isoform X1	0.7612
H2L309	Keratin type Leytoskalatal 18 like	0.7008
C/DAW7	Microsomal apoxida hydrolasa	0.7579
C4rAW/	Fumerata hydrotasa, mitaahandrial lika	0.7329
I3KIC9 I3ID03	40S ribosomal protain S25 like	0.7439
M/ANEQ	Glutamata dahudroganasa, mitashandrial lika	0.718
M4ANE0	Ubiquitin like modifier activating anguma 1	0.7107
A0A0002139	Valosin containing protoin (Fragment)	0.0808
Q_{2}	Chucosa 6 phospheta translocasa isoform V1	0.0002
LINUAJ LINUNUS	Tubulin bota 1 chain	0.0311
$C_{2}OOV_{2}$	Clathrin beauty chain 1 isoform V2	0.0074
030918	rueleelin isoform X2	0.5394
Q43/90 E6V715	Aldehude dehudrogenese femily 8 member Al isoformV1	0.5517
	Aldenyde denydrogenase family 8 member Al isoloimAl	0.5194
Κ/ΓΓD9	Denta-1-pyrtonne-3-carboxyrate denydrogenase, nntochondriar	0.3043
Decreased protein		
EETEuseu proiei EETUUN	Cotachol O mothyltransforaça domain containing protain 1	0 5385
C2NDLO	Transkatalasa	-0.5385
UJINNII UJINNII	Malic onzumo	-0.5545
$\frac{112}{00}$	Cytochroma P450 2P11	-0.5545
	Boto actin 1	-0.5905
L7L0JJ	Dela delle-i	-0.3907
C2DTV7	Endonlesmic ratioulum resident protoin 27	-0.0003
D3TIK0	Alpha amylasa	-0.0394
W5N025	Appla-allylast Protain disulfide isomerose (Fragment)	-0.0403
W JIN 723	Provisional 2.4 dianovil CoA reductors like	-0./119
1V14AA9U MODO A 9	Cotalago	-0./139
MISLAV URLADO	Clutations S. transforaça (Fragment)	-0./390
2003D8 10TV77	Corboxymontidoso A 1 like	-0.0419
11211//		-0.0024

Q8JHC5	Metallothionein (Fragment)	-0.8796
F1Q6E1	4-hydroxyphenylpyruvate dioxygenase	-0.8941
M4AAN9	Phosphate carrier protein, mitochondrial-like isoform X1	-0.9372
K4GAL6	Adenosylhomocysteinase	-0.9525
F7DQ24	11-cis retinol dehydrogenase-like, partial	-0.9856
W5LDH9	Uricase	-1.0609
R0LYE9	Maleylacetoacetate isomerase (Fragment)	-1.0864
F7FYK5	40S ribosomal protein SA-like	-1.1605
B5X3S0	Estradiol 17-beta-dehydrogenase 12-B	-1.1615
M9P0N9	Heart-type fatty acid binding protein	-1.1928
A0A060YQH0	Aspartate aminotransferase, cytoplasmic-like	-1.2149
S9XSM0	Actin, cytoplasmic 2	-1.26
P81399	Fatty acid-binding protein 1, liver	-1.336
H2VEH5	Peptidyl-prolyl cis-trans isomerase	-1.3479
B5X8Y0	Cofilin-2	-1.7414
	ALL	

	ACCEPTED MANUSCRIPT
391	
392	Figure captions
393	Fig. 1.
394	OPLS-DA score plot based on the NSAF values of liver proteins observed during the cooling phase
395	(t1/t0). Cross-validation parameters R2X (cum), R2Y (cum) and Q2 (cum) are reported. The ellipse
396	represents T2 Hotelling's plot with 95% confidence.
397	
398	Fig. 2.
399	OPLS-DA score plot based on the NSAF values of liver proteins observed during the cold maintenance
400	phase (t2/t1). Cross-validation parameters R2X (cum), R2Y (cum) and Q2 (cum) are reported. The ellipse
401	represents T2 Hotelling's plots with 95% confidence.
402	
403	Fig. 3.
404	OPLS-DA score plot based on on the NSAF values of liver proteins undergoing abundance changes
405	along the cooling trial (t2/t0). Cross-validation parameters R2X (cum), R2Y (cum) and Q2 (cum) are

reported. The ellipse represents T2 Hotelling's plots with 95% confidence. 406 x

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Highlights

- ✓ Exposure to cold temperature modifies the sea bream liver protein abundance profile
- ✓ Proteolysis and aminoacid catabolism are most affected during temperature decrease
- ✓ Methionine cycle and sugar metabolism are most affected upon prolonged cold stress
- ✓ Shotgun proteomics complements the data on hepatic metabolism changes following cold stress