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## Runs of homozygosity provide a genome landscape picture of inbreeding and genetic history of European autochthonous and commercial pig breeds

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### **Animal Genetics**

Runs of homozygosity provide a genome landscape picture of inbreeding and genetic history of European autochthonous and commercial pig breeds

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#### **Summary**

Runs of homozygosity (ROH) are long stretches of DNA homozygous at each polymorphic position. The proportion of genome covered by ROH and their length are indicators of the level and origin of inbreeding. Frequent common ROH within the same population define ROH islands and indicate hotspot of selection. In this work, we investigated ROH in a total of 1131 pigs, from 20 European local pig breeds and in three cosmopolitan breeds, genotyped with the GGP Porcine HD Genomic Profiler. PLINK software was used to identify ROH. Size classes and genomic inbreeding parameters were evaluated. ROH Islands were defined by evaluating different thresholds of homozygous SNP frequency. A functional overview of breed-specific ROH islands was obtained via over-representation analyses of Gene Ontology biological processes. Mora Romagnola and Turopolie breeds had the largest proportion of genome covered with ROH (~1003 and ~955 Mb, respectively) whereas Nero Siciliano and Sarda breeds had the lowest proportion (~207 and 247 Mb, respectively). The highest proportion of long ROH (>16 Mb) was in Apulo-Calabrese, Mora Romagnola e Casertana. The largest number of ROH islands was identified in the Italian Landrace (n. 32), Cinta Senese (n. 26) and Lithuanian White Old Type (n. 22) breeds. Several ROH islands were in regions encompassing genes known to affect morphological traits. Comparative ROH structure analysis among breeds indicted similar genetic structure of local breeds across Europe. This study contributed to understand the genetic history of the investigated pig breeds and provided information to manage these pig genetic resources. 

Keywords: Autozygosity; Population genomics; Selection signature; SNP; Sus scrofa

### Introduction

Conservation programs of animal genetic resources, mainly constituted by numerous autochthonous breeds in all species, are usually challenged by their very small effective population size which, in turn, tends to increase inbreeding and to reduce genetic variability (Charlesworth & Willis 2009). Inbreeding depression is considered the result of the increased level of autozygosity. Pedigree information is traditionally used to calculate the inbreeding coefficient (F<sub>PED</sub>), defined as the probability that in a diploid individual, the maternal and the paternal derived alleles at a randomly selected locus are identical by descent (Wright 1922). This definition is equivalent to consider F<sub>PED</sub> as the proportion of autozygosity of an individual's genome. Then, the level of inbreeding of a population is expressed by averaging all F<sub>PED</sub> individual values. Reliability of F<sub>PED</sub> calculated in autochthonous breeds is in general lower than what is possible to obtain for animals in commercial selection nuclei. This is mainly due to incomplete registration and incorrect recording of all mating events derived by the extensive production systems in which local breeds are usually raised (Gomez-Raya et al. 2008; Kios et al. 2012). In addition, it is clear that a few assumptions used to calculate this pedigree-based coefficient are not correct and are used as approximations in the methods of calculations: i) all founder animals of the base population are expected to be unrelated, but this condition cannot be evaluated and it is usually not respected; ii) recombinant events occurring during meiosis mix equally the individual's paternal and maternal haploid genome copies, but this condition mimics only average events and not what actually happens in each specific meiosis; and iii) there is no selection biases on any parts of the genome, but this assumption is not respected considering that directional artificial selection or natural selection play important roles in shaping the genome of many domestic animal breeds. 

Genome wide analyses, usually based on single nucleotide polymorphism (SNP) arrays, can be used to estimate the level of autozygosity of an animal genome by directly interrogating the genotype status at thousands of polymorphic sites (e.g. Kristensen *et al.* 2010). The proportion of the genome covered by runs of homozygosity (ROH) of a certain minimal length has been considered one of the

most precise estimation of the level of autozygosity, providing a measure of genomic inbreeding ( $F_{ROH}$ ; Peripolli *et al.* 2017). Runs of homozygosity are defined as continuous chromosome stretches in which all loci have a homozygous genotype (Gibson *et al.* 2006). Some ROH characteristics in a population (the average length of ROH, the average proportion of the genome covered by ROH and the patterns of ROH distribution across the chromosomes) are considered indicators of the origin and genetic history of a population (Ceballos *et al.* 2018). The high frequency of ROH in some chromosome regions identifies selection signatures derived from a reduced haplotype variability around loci under natural or artificial selection (i.e. ROH island or ROH hotspots). By applying different strategies and methods, ROH islands have been used to detect signatures of selection in several livestock species (Purfield *et al.* 2017; Bertolini *et al.* 2018; Grilz-Seger *et al.* 2018; Mastrangelo *et al.* 2018; Peripolli *et al.* 2018), including the pig (Zhang *et al.* 2018; Gorssen et al. 2020; Schiavo *et al.* 2020b).

A lot of different pig breeds have been developed through the combined action of artificial directional selection and natural pressures that contributed to shape a large reservoir of genetic diversity within the Sus scrofa species (Porter 1993). A large fraction of these genetic resources is however constituted by autochthonous breeds of small population size, usually well adapted to their local agro-climatic and environmental conditions but less productive, compared to cosmopolitan breeds or lines. Conservation programmes for these breeds, some of which considered unexplored genetic resources, have different levels of managing actions that range from advanced Herd Book structures with specific breeding and selection plans to preliminary voluntary farmer-based herd books or primitive conservation programmes (Čandek-Potokar & Nieto 2019). We recently analysed major and candidate gene markers in 20 autochthonous European pig breeds from several different countries and obtained preliminary population structure results (Muñoz et al. 2018) that were refined using SNP array information (Muñoz et al. 2019) and whole genome resequencing data (Bovo et al. 2020a, 2020b). Genome wide data indicated that average persistence and strength of linkage disequilibrium between markers and SNP based effective population size varied among breeds

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depending by the genetic structures and history of these breeds that experienced different genetic events (e.g. admixture, bottlenecks and genetic drift). Selection signatures were also obtained using  $F_{ST}$  statistics by analyzing SNP chip genotyping and sequencing data (Muñoz *et al.* 2019; Bovo *et al.* 2020a). Genomic inbreeding analyses in these breeds could add other information to refine their conservation programmes and identify appropriate strategies to control inbreeding level and infer other population structures or features.

In this study we analysed the same 20 European autochthonous pig breeds from nine different countries (Croatia, France, Germany, Italy, Lithuania, Portugal, Serbia, Slovenia and Spain) and other three cosmopolitan-derived breeds to obtain genomic inbreeding information from whole genotyping datasets by using ROH and other genomic approaches. We then evaluated the distribution of ROH in the genome of these breeds and identified putative selection hotspot regions that might be originated by different selection histories and structures of these pig genetic resources.

### **Materials and methods**

### 2 Animals

Pigs included in this study were from 20 autochthonous breeds distributed in nine European countries (Alentejana and Bísara from Portugal; Iberian and Majorcan Black from Spain; Basque and Gascon from France; Apulo-Calabrese, Casertana, Cinta Senese, Mora Romagnola, Nero Siciliano and Sarda from Italy; Krškopolje from Slovenia; Black Slavonian and Turopolje from Croatia; Moravka and Swallow-Bellied Mangalitsa from Serbia; Schwäbisch-Hällisches Schwein from Germany; Lithuanian indigenous wattle and Lithuanian White old type from Lithuania) and three commercial breeds (Italian large White, Italian Landrace and Italian Duroc). Analysed pigs were selected by avoiding highly related animals (no full- or half-sibs). All animals had standard breed characteristics and were registered to their respective Herd Books. Table S1 reports detailed descriptions of the investigated breeds and selected animals (Čandek-Potokar & Nieto 2019). Pictures

of animals of the autochthonous breeds are reported in Muñoz et al. (2018, 2019) and Bovo et al. 143 (2020a). 144

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#### 10 146 Genotyping of single nucleotide polymorphisms

12 All pigs (39-55 for each breed; Table S2) were genotyped with the GeneSeek ® GGP Porcine 147 HD Genomic Profiler v1 (Illumina Inc, USA), which includes 68,516 SNPs evenly distributed with 15 148 17 149 a median of 25 kb gap spacing. The average genotyping call rate was 0.94. Single nucleotide polymorphisms were mapped on the Sscrofa11.1 genome version, following the procedure already 150 described (Fontanesi et al. 2012, 2014). Only autosomal SNPs located in unique positions were 22 <sup>151</sup> 24 152 considered. Genotyping data were then filtered using PLINK software version 1.9 (Chang et al. <sup>26</sup> 153 2015). Call rate of 0.90 and Hardy Weinberg equilibrium P of 0.001 were set as thresholds to keep 154 SNPs. Although filtering for minor allele frequency (MAF) is necessary as best practice in most SNP 31 155 chip analyses, this approach excludes the SNPs that are homozygous for the whole breed, therefore 33 156 it could bring to an underestimation of the coverage in ROH (Meyermans et al. 2020). For this reason, we analysed ROH without applying any MAF pruning. For comparison with other studies that applied 157 <sub>38</sub> 158 a MAF threshold and to evaluate the impact of MAF on the calculated ROH parameters, we also used a MAF threshold of 0.01 (indicated as method based on MAF > 0.01) and results are included in the 40 159 42 160 Supplementary material. All analyses in the text are derived without MAF pruning (indicated as 45 161 method based on MAF  $\geq$  0.00), if not stated otherwise. Animals were discarded if their call rate was <0.90. Table S2 reports the number of SNPs and animals considered for further analyses after 47 162 163 filtering.

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## Multidimentional-plot analysis of pig breeds and effective population size

The first three dimensions for a multidimensional (MDS)-plot have been obtained with PLINK software version 1.9 and plotted with the R package "Scatterplot3d" (Ligges & Mächler 2003) to 167 168 graphically visualize the genetic distances between the 23 pig breeds. Effective population size at

recent and remote generations was computed using SNP data with the software SNeP (Barbato et al. 169 2015) with default parameters, except for the maximum distance in bp between SNPs to be analysed, 170 that has been set to 10 Mb, and the binwidth for the calculation of linkage disequilibrium that was set 171 10 172 to 100 kb.

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# Identification of runs of homozygosity

Runs of homozygosity (ROH) were identified using PLINK software version 1.9 (Chang et al. 17 175 2015). No pruning was performed based on linkage disequilibrium to avoid biases that could be 176 derived by this practice (Marras et al. 2015; Meyermans et al. 2020) but a minimum length of 1 Mb was set to detect ROH. This threshold may exclude short and common ROH determined by markers in linkage disequilibrium, as previously demonstrated (e.g. Ferencakovic et al. 2013; Marras et al. 2015). The following parameters, already used by Schiavo et al. (2020a), were considered to call 180 ROH: i) the minimum number of consecutive homozygous SNPs included in the ROH was 15; ii) the minimum length that constituted the ROH was 1 Mb; iii) the number of heterozygous SNPs that were allowed in the ROH was 0; iv) the minimum density of SNP in a genome window was 1 SNP every 183 100 kb; v) the maximum gap between consecutive SNPs was 1000 kb. ROH were placed into five size classes (Kirin et al. 2010; Ferenčaković et al. 2013a; Schiavo et al. 2020a): 1-2, 2-4, 4-8, 8-16 186 and >16 Mb, identified as ROH1-2 Mb, ROH2-4 Mb, ROH4-8 Mb, ROH8-16 Mb and ROH>16 Mb, respectively. The total number of ROH (nROH) was then obtained for each individual and for each length class. The average length of ROH (L<sub>ROH</sub>, in Mb) and the sum of all ROH segments by 189 animals (S<sub>ROH</sub>, in Mb) were calculated. These parameters were also calculated for each breed by 190 averaging individual data.

### Genomic inbreeding measures

193  $F_{ROH}$  was calculated for each pig as the proportion of the autosomal genome covered by ROH. 194  $F_{ROH}$  was calculated using all the detected ROH with length >1 Mb ( $F_{ROH1}$ ) and also considering Page 9 of 36

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higher thresholds of length, namely >4 Mb, >8 Mb, >16 Mb to obtain, respectively,  $F_{ROH4}$ ,  $F_{ROH8}$  and F<sub>ROH16</sub> inbreeding coefficients. Averaged  $F_{ROH}$  values were calculated for each breed. In addition, chromosome (SSC)  $F_{ROH}$  ( $F_{ROHSSC}$ ) values were also estimated for each breed:  $F_{ROHSSC} = L_{ROHSSC}/L_{SSC}$  (Silió *et al.* 2013), in which  $L_{ROHSSC}$  is the total length of an individual's ROH in each SSC and  $L_{SSC}$  is the length of each chromosome covered by the involved SNPs.

Other genomic inbreeding coefficients were calculated: i) the variance-standardized relationship minus 1 ( $F_{hat1}$ ); ii) the excess of homozygosity-based inbreeding estimate ( $F_{hat2}$ ); iii) the estimate based on correlation between uniting gametes ( $F_{hat3}$ ); iv) the values of the diagonal elements of the genomic relationship matrix, GRM ( $F_{GRM}$ ; Van Raden *et al.* 2011); v) the difference between observed and expected number of homozygous genotypes ( $F_{HOM}$ ).  $F_{hat1}$ ,  $F_{hat2}$ ,  $F_{hat3}$  and  $F_{GRM}$ . GRM coefficients were calculated using PLINK1.9 with the ported functions of GCTA software v. 1.92 (Yang *et al.* 2011).  $F_{HOM}$  was computed with PLINK software version 1.9 (Chang *et al.* 2015). Pearson correlation coefficients (r) between all evaluated inbreeding coefficients were calculated.

### Identification of runs of homozygosity islands and annotation of genome regions

First, the proportion of SNPs residing within a ROH was calculated for a given breed by counting the amount of times a SNP appeared in a ROH within the given breed divided by the total number of genotyped pigs of that breed. Then, to call ROH islands a threshold of frequency should be defined. A few methods have been proposed for this purpose, each with pros and cons (e.g. Purfield *et al.* 2017; Grilz-Seger *et al.* 2018, Gorssen *et al.* 2020). However, there is no general agreement on their use in different contexts and populations. In this study, we used three methods to identify ROH islands that differed on the threshold that was applied.

One method already reported in other studies (Grilz-Seger *et al.* 2018, 2019a, 2019b) uses an empirical threshold defined as the percentage of animals (usually 50%), whiting a population, positive for a ROH at each tested SNP (hereinafter called 50% of animals-based threshold). When the level of inbreeding is high, the identification of islands due to signature of selection based on a fixed

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percentage of animals having ROH at each position of the genome might increase the number of false positive ROH islands that indicate the presence of signature of selection. This method could increase the risk of type II errors when the level of inbreeding in the population is low. Another method, frequently applied for this aim (e.g. Szmatoła *et al.* 2016; Purfield *et al.* 2017; Bertolini *et al.* 2018; Mastrangelo *et al.* 2018; Zhang *et al.* 2018), defines a percentile threshold (99th percentile) based on the top 1% of SNPs observed in a ROH in each breed (hereinafter called percentile-based threshold). Adjacent SNPs over this threshold are then merged into genomic regions corresponding to ROH hotspots. This method identifies always ROH islands as the threshold is defined on a percentile within the breed dataset and does not consider the structure of the population or its level of inbreeding.

Considering the problems that these two methods could have, we developed a third method where the identification of the threshold was chosen using a linear model in which the number of animals having SNPs in a ROH was a function of the average  $S_{ROH}$  level of the breed, which approximate the genomic inbreeding level of a population (hereinafter called  $S_{ROH}$  based-threshold). ROH islands were then considered in the text and annotated based on the results derived by this latter method. Results obtained with the other two methods were used for a comparative analysis. ROH cooccurrence between different breeds were investigated by comparing the average homozygosity level in each breed at each island region. For this evaluation, each ROH island identified in at least one breed was considered.

Similarity among breeds was investigated by computing a first matrix **A** (*n* breeds × *m* ROH islands regions identified across all the analyzed breeds) whose generic entry *a* is the average breedspecific frequency value of a given ROH island computed as follows:  $a = \frac{\sum_i AF_i}{n}$ , where AF<sub>i</sub> is the allele frequency of the i<sup>th</sup> SNP belonging to the ROH island and including *n* SNPs. This matrix was used to compute a similarity matrix D (*n* × *n*), whose generic entry *d* is the Euclidean distance between pairs of breeds with values scaled in the range 0 to 1. A final dissimilarity matrix (1-D) was obtained

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and used to produce a heatmap in R (package *corrplot*; Wei and Simko, 2007) showing similarity
among breeds.

Genes annotated in the Sscrofa11.1 pig genome version that mapped in the identified ROH islands were retrieved using the Ensembl Biomart tool (http://www.ensembl.org/biomart/martview/) and from NCBI Sscrofa11.1 GFF file. Functional enrichment analysis was carried out with *Enrichr* (Chen *et al.* 2013) via Fisher's exact test. Analyses run over the Biological Process (BP) branch of the Gene Ontology (GO) (Ashburner *et al.* 2000), by interrogating a total of 5103 functional terms covering 14433 human genes. Breed-specific analyses were run by using as input set the list of genes included in ROH islands. We considered as statistically over-represented terms those having: i) at least two input genes from two or more different ROH islands and ii) an adjusted *P* lower than 0.10.

### Results

## Genomic relationships among breeds and effective population size

Genomic information on the analysed breeds based on SNP data was graphically presented in a tri-dimensional MDS-plot (Figure S1). This plot showed that distinct groups of individuals were usually from the same breed. Several breeds were well separated from other groups. These distinct groups included breeds from several countries: Gascon and Basque from France; Italian Large White, Italian Duroc and Mora Romagnola from Italy; Iberian from Spain; Turopolje from Croatia. Most of the other breeds formed a continuous large cluster showing a general geographical distribution gradient as already reported in principal component analyses that included the same autochthonous breeds (Muñoz *et al.* 2019).

Effective population size (*Ne*) estimated with software *SNeP* for the 23 breeds is reported in Table S3. For all 20 autochthonous breeds, results confirmed the general low *Ne* for most breeds as already reported by Muñoz *et al.* (2019) who applied a similar estimation method. At 5 generations ago, breeds with the lowest *Ne* values were Turopolje, Mora Romagnola, Apulo-Calabrese and Casertana (*Ne* = 15, 16, 22 and 22, respectively). These breeds had the lowest estimated *Ne* also in

the study of Muñoz *et al.* (2019) even if in different order. The autochthonous breeds with the largest *Ne* were Iberian, Nero Siciliano, Alentejana, Majorcan Black, Sarda and Bísara (Ne = 69, 68, 61, 58, 57 and 55, respectively). The commercial breeds had a higher *Ne* than all other remaining autochthonous breeds. In Italian Duroc, Italian Landrace and Italian Large White  $N_e$  at 5 generation ago was equal to 53, 59 and 61, respectively.

### Runs of homozygosity in the investigated breeds

Table 1 (MAF  $\ge$  0.00) and Table S4 (MAF > 0.01) show the average size and average number of ROH (considering all ROH>1 Mb) per pig (average L<sub>ROH</sub> and average nROH, respectively) and the average S<sub>ROH</sub> values per animal in the 23 breeds. Minimum and maximum values for these three parameters are reported in Table S5. As expected, the parameters calculated without any MAF pruning were always higher than the parameters calculated using MAF >0.01. The breeds that had the highest mean nROH were Basque, Italian Duroc and Turopolje (n. 107, n. 104 and n. 80, respectively) and the breeds with the lowest mean nROH were Nero Siciliano (n. 24) Sarda (n. 27) and Moravka (n. 30). The mean L<sub>ROH</sub> in all autochthonous breeds was larger than that of all three commercial breeds. Three Italian local breeds (Mora Romagnola, Apulo-Calabrese, and Casertana had the largest L<sub>ROH</sub> (14.38, 14.21 and 12.63 Mb, respectively). Among the autochthonous breeds, the lowest L<sub>ROH</sub> was observed in Alentejana (6.49 Mb), Iberian (6.50 Mb) and Majorcan Black (6.58 Mb). The maximum ROH length was observed in the largest chromosomes and reached 24.34 Mb in Mora Romagnola (SSC1), 23.36 Mb in Nero Siciliano (SSC1), 22.64 Mb in Moravka (SSC1) and 21.55 Mb in Apulo-Calabrese (SSC13). Mora Romagnola and Turopolje breeds had the largest mean S<sub>ROH</sub> (a total of ~1003 and ~955 Mb, respectively) whereas Nero Siciliano and Sarda breeds had the lowest mean values for this parameter (~207 and ~247 Mb, respectively). The maximum  $S_{ROH}$  value was observed in one Mora Romagnola and one Black Slavonian pig that had about half of their genome covered by ROH (Table S5).

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Figure 1 shows the correlation plots between the  $S_{ROH}$  and the nROH values over the individual pigs in the 23 breeds. Basque and Gascon showed very homogeneous plots, indicating that most pigs of these two breeds had very similar within individual ROH parameters (nROH,  $L_{ROH}$  and  $S_{ROH}$ ). The opposite was the heterogeneous distribution observed in the Apulo-Calabrese, Bísara, Casertana and Turopolje breeds (Figure 1).

Figure 2 reports the proportion of ROH of the five different length classes in each breed. Table S6 lists the corresponding values. The highest proportion of long ROH (>16 Mb) was in Apulo-Calabrese, Mora Romagnola e Casertana (about 25%, 23% and 23%, respectively). Apulo-Calabrese, Casertana, Mora Romagnola and Turopolje had the lowest proportion of short-medium ROH (ROH1-8). All three commercial breeds, Alentejana, Gascon, Iberian, Majorcan Black, Nero Siciliano, Lithuanian indigenous wattle, Lithuanian White Old Type and Schwäbisch-Hällisches had more than 50% of short ROH (ROH1-2 and ROH2-4).

### 09 *Genomic inbreeding parameters based on runs of homozygosity*

Table 2 reports the mean and standard deviation of genomic inbreeding parameters calculated using ROH from different size classes in the 23 breeds. Mora Romagnola, Turopolje and Apulo Calabrese and Casertana were the autochthonous breeds with the highest  $F_{ROH}$  values, considering all ROH classes. For example, among these breeds  $F_{ROH1}$  ranged from 0.409 (Mora Romagnola) to 0.243 (Casertana). Among the commercial breeds, Italian Duroc had the highest  $F_{ROH}$  values. The lowest  $F_{ROH1}$  levels were observed in Nero Siciliano (0.085), Sarda (0.101) and Moravka (0.118).

When considering only medium-long ROH to calculate other ROH based inbreeding parameters (i.e.  $F_{ROH4}$ ,  $F_{ROH8}$  and  $F_{ROH16}$ ), the values decreased in all breeds, as expected. Among those with high  $F_{ROH1}$ , this drop was more evident in the breeds that had a high percentage of short ROH than in breeds that had many long ROH. For example, the Italian Duroc  $F_{ROH16}$  value was about 2.5 times lower than that of  $F_{ROH1}$  value whereas in Mora Romagnola, Turopolje, Apulo-Calabrese and Casertana their  $F_{ROH16}$  values decreased only 1.4-1.6 times compared to their respective  $F_{ROH1}$ 

values. The distribution of the F<sub>ROH</sub> values in the analysed breeds is shown in the boxplots of Figure
3.

The genome wide F<sub>ROH</sub> information was also dissected by considering the average proportion of all ROH covering the different autosomes (F<sub>ROHSSC</sub>). Among all breeds, Mora Romagnola and Turopolje had the highest F<sub>ROHSSC</sub> values for 10 (SSC1, SSC4, SSC8, SSC9, SSC10, SSC13, SSC14, SSC15, SSC16 and SSC17) and 5 (SSC2, SSC3, SSC5, SSC6 and SSC11) chromosomes, respectively. Apulo-Calabrese had the highest F<sub>ROHSSC</sub> values for SSC7 and SSC18 whereas Basque had the highest F<sub>ROHSSC</sub> value for SSC12 (Figure S2).

Mean  $F_{ROH1}$ ,  $F_{ROH4}$ ,  $F_{ROH8}$  and  $F_{ROH16}$  breed values were negatively correlated with the estimated breed *Ne* values at 5 generation ago, defined as reported above (r = -0.685, -0.722, -0.737 and -0.716, respectively; P <0.0001).

### Other genomic inbreeding parameters and their correlations with $F_{ROH}$

Other parameters that have been proposed as estimators of the level of genomic inbreeding were calculated in the 23 breeds (Table S8). The average  $F_{hat1}$  value was positive in only two breeds (Apulo-Calabrese and Sarda) and ranged from -0.320 (Mora Romagnola) to 0.010 (Sarda), with large within breed variability (the largest standard deviation was in Turopolje) and among breeds variability. These considerations could be also applied for the  $F_{GRM}$  parameter which is equivalent to  $F_{hat1}$  (even if scaled in a different way). The average  $F_{hat2}$  and  $F_{hat3}$  parameters had both the extreme values for the same breeds (Lithuanian indigenous wattle with the lowest values and Apulo-Calabrese with the highest values) with similar within and among breed variability (Table S8). The average  $F_{HOM}$  values were negative in 11 out or 23 breeds and ranged from -0.070 in Lithuanian Indigenous Wattle to 0.124 in Apulo-Calabrese. Turopolje had the largest standard deviation for this parameter (0.24). Distribution plots of the  $F_{hat1}$ ,  $F_{hat2}$ ,  $F_{hat3}$  and  $F_{HOM}$ , parameters in the analysed breeds are reported in Figure S3 and Figure S4.

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Correlations between all  $F_{ROH}$  parameters and all other genomic inbreeding measures for each breed are reported in Table S9.  $F_{HOM}$  had always very high and consistent correlations with the ROH based measures over all breeds. For example, correlations with  $F_{ROH1}$  and  $F_{ROH4}$  ranged from 0.819 and 0.814 for the Nero Siciliano breed to 0.987 and 0.982 for the Bisara breed. Correlations between  $F_{hat2}$  and  $F_{ROH1}$  and  $F_{ROH4}$  had some lower values even if again very high and consistent across breeds (they ranged from 0.447 or 0.450 in Swallow-Bellied Mangalitsa to 0.909 and 0.906 in Casertana).  $F_{hat1}$  and  $F_{hat3}$  showed inconsistent correlations compared to those of the other measures, including also negative values (Table S9). All these other genomic inbreeding measures had low negative correlations with *Ne* (from -0.11 to -0.18).

57 Run of homozygosity islands

Table 3 summarizes the number of ROH islands and the fraction of the genome covered by ROH islands identified using the S<sub>ROH</sub> based-threshold in the 23 pig breeds. Figure 4 includes the Manhattan plots of a few breeds with extreme numbers of ROH islands. Figure 5 reports the pairwise similarities between breeds when overlapping ROH islands across breeds were considered. Some common features across breeds were evident.

The largest number of ROH islands was identified in the Italian Landrace (n. 34), Cinta Senese (n. 26) and Lithuanian White Old Type (n. 22) breeds. The largest covered fraction of the genome was observed in the Italian Duroc (92.85 Mb), Turopolje (80.82 Mb, with the largest averaged size of ROH islands) and Italian Landrace (75.03 Mb). No ROH islands were observed in Apulo-Calabrese and in Sarda breeds.

Table S10 compares the results obtained using the  $S_{ROH}$  based-threshold method with the results obtained using the other two methods considered in this study (the 50% of animals-based threshold and the percentile-based threshold methods, see Materials and methods). The Manhattan plots for all breeds and including the thresholds derived by the three methods is reported in Figure S5. Breeds with the highest level of genomic inbreeding estimated using  $F_{ROH}$  measures, like Mora Romagnola,

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Turopolje and Basque (Table 2), showed the highest number of ROH islands and the largest fraction of genome covered by ROH islands with the 50% of animals-based threshold method (n. 91 with 756 Mb in Mora Romagnola, n. 129 with 747 Mb in Turopolje and n. 93 in Basque with 312.9 Mb). Using the percentile-based threshold method, the number of ROH islands and the total length of the genome fractions covered by these regions were similar in all breeds and ranged from n. 7 (Mora Romagnola) to n. 20 (Italian Landrace ) and from 19.83 Mb (Casertana) to 44.51 Mb (Turopolje). These methods could capture different information from the analysed populations. It seems however, that these two latter methods are, to some extent, biased by the genetic structure of the analysed populations and by the methodologies that are applied.

The complete list of ROH islands identified in the investigated breeds, using the S<sub>ROH</sub> basedthreshold method, including the genes annotated in these regions, is reported in Table S11. Several breeds had ROH islands encompassing genes that are well known to affect exterior traits, that might contribute to differentiate these pig breeds. For example, Gascon and Turopolje had a ROH island on SSC6 which includes the *melanocortin 1 receptor* (*MCIR*) gene and Krškopolje and Turopolje had another ROH island on SSC8 which includes the *v-kit Hardy-Zuckerman 4 feline sarcoma viral oncogene homolog* (*KIT*) gene. These two genes are well known to affect coat colour and colour patterns (Fontanesi & Russo 2013). Two genes that are known to affect vertebral number (*nuclear receptor subfamily 6 group A member 1*, *NR6A1* on SSC1; and *vertnin*, *VRTN* on SSC7; Mikawa *et al.* 2007, 2011) were in two ROH islands observed in Italian Landrace and in Schwäbisch-Hällisches breeds, respectively. Moravka and Schwäbisch-Hällisches breeds had a ROH island on SSC5 including the *methionine sulfoxide reductase B3* (*MSRB3*) gene whose variants have been associated with ear size in pigs (Chen *et al.* 2018; Bovo *et al.* 2020a). Cinta Senese and Italian Duroc had a ROH island including other genes that have been shown to affect body size (*caspase 10*, *CASP10*; and *non-SMC condensin I complex subunit G*, *NCAPG*; Rubin *et al.* 2012).

A functional overview of breed-specific ROH islands identified using the  $S_{ROH}$  based-threshold method was obtained via over-representation analyses of GO biological processes (Table S12). Few Page 17 of 36

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terms characterizing ROH islands were detected in two breeds (Krškopolje and Swallow-Bellied Mangalitsa) only. Terms were general and included pattern recognition receptor signaling pathway, toll-like receptor signaling pathway, zymogen activation, cellular response to radiation and negative regulation of cell differentiation.

### Discussion

The demographic history of a population can be inferred using information from the average distribution, coverage, size and patterns of ROH that can be identified in the individuals belonging to the population using high density SNPs data (Ceballos et al. 2018). In this study we detected ROH in the genome of pigs from 20 autochthonous and three commercial breeds and compared the obtained ROH genome landscapes patterns. These breeds represent populations that derived from several countries and originated in different production systems that largely contributed to shape their genetic structures.

Combining different population genomic parameters calculated in this study it could be possible to reconstruct, to some extent, the genetic events and history that contributed to define the current genetic pools of the investigated breeds. ROH based fingerprinting are left in the analysed breeds and can be used to divide the 23 breeds in a few macro-groups that could have independently experienced similar genetic trajectories.

The ROH complement of recently inbred populations is defined by a large number of ROH with large size and a large fraction of the genome covered by ROH (high  $S_{ROH}$ ), owing to very recent pedigree inbreeding loops, accompanied by a small Ne. The large  $S_{ROH}$  standard deviation indicates a low uniformity of the animals, that means that there might be different substructures or heterogeneity in the population or that an original bottleneck or founder effect could have increased the range of ROH size. Recent inbreeding features accompanied by a constituting bottleneck series of events can be clearly evidenced in a few Italian local breeds, i.e. Apulo-Calabrese, Casertana, Mora Romagnola, and in Turopolie. The high level of inbreeding could have masked regions that harbor 

selection of signatures as most of these breeds showed a low number of ROH islands (from zero to 7, considering the  $S_{ROH}$  based method; Table 3) apart Turopolje that seems to maintain a quite high level of ROH specific regions (n. 17; Table 3). These breeds need to be carefully managed to reduce or control the high level of inbreeding. Programmes in this direction are currently under way in the Italian breeds (ANAS, 2020).

Other breeds have a quite high S<sub>ROH</sub> level but with short ROH indicating the occurrence of a past bottleneck and then a quite good isolation of the genetic pool. This is a case that can be observed in the two French breeds, Basque and Gascon, and in the Italian breed Cinta Senese. Differences in the three breeds are evident in the number of ROH islands that might indicate a low-medium level of specific signatures of selection in the French breeds (7 in the Basque that also had the largest number of nROH among the three - and 12 in the Gascon) and a high level of characterizing signatures in the Cinta Senese (26 ROH islands) probably due to different levels of selection pressures and adaptation of the three considered populations. A similar genetic history seems evident in the Italian Duroc breed (which however had a larger *Ne*; Table S3), reflecting deeper parental relatedness and consistent with an original strong bottleneck that occurred at the beginning of the 1990' when the heavy pig selection programme was defined and differentiated the Italian Duroc breed from other Duroc lines (Bosi & Russo 2004).

Breeds that experienced recent admixtures had, in general, a low nROH and as a proportion, had a higher frequency of short-medium ROH than long ROH, with high *Ne*. This group included the two breeds that had nROH <30,  $S_{ROH}$  <300.00 Mb, and Ne >55, i.e. Nero Siciliano and Sarda for which the ROH derived landscape was in agreement with the large variability observed in candidate gene markers and SNP chip data (Muñoz *et al.* 2018, 2019). Other breeds (i.e. Alentejana, Black Slavonian, Krskopolje, Lithuanian indigenous wattle and Moravka) had similar ROH patterns with that described for these two Italian breeds even if not so extreme (nROH <40, S<sub>ROH</sub> <350.00 Mb). They constitute a heterogeneous group of populations that might have experienced some moderate introgression over the period of their constitution or that these events occurred in the past and at

present they maintain a moderate level of variability. The low-medium number of ROH islands (from 3, Moravka, to 15 Krskopolje) indicates a low-medium level of differentiation in terms of specific ROH features. Another group of intermediate breeds (which some features partially overlapping with those of the previous group) with medium nROH and, in general, with a medium level of inbreeding (nROH>40 and  $S_{ROH}>300$ ) includes Bísara, Lithuanian White Old Type, Majorcan Black, Schwäbisch-Hällisches and Swallow-Bellied Mangalitsa.

Three other breeds, i.e. Iberian, Italian Landrace and Italian Large White, had characteristic ROH derived feature of commercial breeds or large populations, as expected from their large population size (consistent with the large *Ne*). The two Italian breeds had some indicators of more specific differentiations and signatures of selection with a higher number of nROH, lower *Ne* and larger fraction of the genome included in ROH islands than the Iberian breed. This fact could be also due to the high level of genetic diversity observed within the Iberian breed, sometime higher than in some European pig breeds (Fabuel *et al.* 2004). This is consistent with the structure of these three populations, with the two Italian breeds being derived by small selection nuclei specifically addressing a selection programme for heavy pigs. The presence of common features among breeds raised in different countries suggests that a few ROH islands might capture some adaptive features that are shared across populations and production systems.

The general picture depicted by the ROH profiles was able to summarize the main elements that characterize the population structure of the analysed breeds. For a few of them the potential burden derived by the ROH should be evaluated with attention. An increased homozygosity for (partially) recessive detrimental mutations maintained at low frequency in populations by mutation– selection balance has been suggested to be one of the main causes of inbreeding depression. Genomic inbreeding measures can help to manage all these pig populations.  $F_{ROH}$  based measures seems more appropriate than all other calculated parameters and are highly correlated with *Ne* indicating that they better reflect the population structure and then the effective inbreeding level of the animals, as we

already reported comparing these measures with pedigree based inbreeding estimations (Schiavo et 476 al. 2020a). 477

The method that we considered to identify ROH islands considers the level of inbreeding of the 478 10 479 breeds to reduce the biases derived by the large fraction of the genome covered by ROH in highly 480 inbreed populations and to increase the probability to capture signatures of selection able to explain 15 481 morphological or adaptative features that characterize the uniqueness of these genetic resources. 17 482 Some of the ROH islands contained genes responsible for domestication signatures related to exterior traits and morphological adaptation (i.e. coat colour genes: MCIR and KIT; Fontanesi & Russo 2013; 483 vertebral number: NR6A1 and VRTN, Mikawa et al. 2007, 2011; parts of the body and body size: 484 24 485 CASP10, MSRB3 and NCAPG; Rubin et al. 2012; Chen et al. 2018) indicating that fixation or 26 486 increased frequency for some haplotypes containing breed specific alleles or features differentiating 487 the domestic pool from wild boars could be captured by ROH.

31 488 Runs of homozygosity can complement other methods that have been applied to extract 33 489 signatures of selection in these pig breeds (Muñoz et al. 2018, 2019; Bovo et al. 2020a, 2020b) and 490 can provide additional information useful to design conservation plans and mating strategies to <sub>38</sub> 491 maintain the diversity of these pig genetic resources.

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#### 56 499 **Conflict of interests**

500 The authors declare they do not have any competing interests.

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5 6 503	Genotyping data of the autochthonous breeds can be shared after the signature of an agreement on
7 8 504 9	their use with the TREASURE Consortium. Genotyping data of the commercial breeds can be shared
9 10 505 11	after the signature of an agreement on their use with the University of Bologna.
<sup>12</sup> 13 506	
14 15 507	References
16 17 508 18	ANAS 2019. Registro Anagrafico. Retrieved on 6th October 2020, from http://www.anas.it/.
<sup>19</sup> 509 20	Ashburner M., Ball C.A., Blake J. A., Botstein D., Butler H., Cherry J. M., Davis A.P., Dolinski K.,
21 22 510	Dwight S. S., Eppig J.T., Harris M. A., Hill D. P., Issel-Tarver L., Kasarskis A., Lewis S., Matese
23 24 511	J.C., Richardson J. E., Ringwald M., Rubin G.M. & Sherlock G. (2000) Gene ontology: Tool
25 26 512 27	for the unification of biology. Nature Genetics 25, 25-9.
28 29 513	Barbato M., Orozco-terWengel P., Tapio M. & Bruford M.W. (2015) SNeP: A tool to estimate trends
30 31 514	in recent effective population size trajectories using genome-wide SNP data. Frontiers in
32 33 515 34	Genetics 6, 109.
$\frac{35}{36}$ 516	Bertolini F., Cardoso T.F., Marras G., Nicolazzi E.L., Rothschild M.F. & Amills M. (2018) Genome-
37 38 517	wide patterns of homozygosity provide clues about the population history and adaptation of
39 40 518	goats. Genetics Selection Evolution 50, 59.
41 42 519 43	Bosi P. & Russo V. (2004) The production of the heavy pig for high quality processed products.
44 45 520	Italian Journal of Animal Science 3, 309–21.
46 47 521	Bovo S., Ribani A., Muñoz M., Alves E., Araújo J.P., Bozzi R., Čandek-Potokar M., Charneca R., Di
48 49 522 50	Palma F., Etherington G., Fernandez A.I., García F., García-Casco J., Karolyi D., Gallo M.,
50 51 52 523	Margeta V., Martins J.M., Mercat M.J., Moscatelli G., Núñez Y., Quintanilla R., Radović Č.,
53 54 524	Razmaite V., Riquet J., Savić R., Schiavo G., Usai G., Utzeri V.J., Zimmer C., Ovilo C.,
55 56 525	Fontanesi L. (2020a) Whole-genome sequencing of European autochthonous and commercial
57 58 59 526	pig breeds allows the detection of signatures of selection for adaptation of genetic resources to
60 527	different breeding and production systems. Genetics Selection Evolution 52, 33.

<sup>3</sup> 528 4	Bovo S., Ribani A., Muñoz M., Alves E., Araújo J.P., Bozzi R., Charneca R., Di Palma F., Etherington
5 6 529	G., Fernandez A.I., García F., García-Casco J., Karolyi D., Gallo M., Gvozdanović K., Martins
7 8 530 9	J.M., Mercat M.J., Núñez Y., Quintanilla R., Radović, Razmaite V., Riquet J., Savić R., Schiavo
10 531 11	G., Škrlep M., Usai G., Utzeri V.J., Zimmer C., Ovilo C. & Fontanesi L. (2020b) Genome-wide
<sup>12</sup> 532 13	detection of copy number variants in European autochthonous and commercial pig breeds by
14 15 533 16	whole-genome sequencing of DNA pools identified breed-characterising copy number states.
17 534 18	Animal Genetics 51, 541–56.
<sup>19</sup> 535 20	Čandek-Potokar M. & Nieto L.R.M. (2019) European Local Pig Breeds - Diversity and Performance.
21 22 536	A study of project TREASURE. IntechOpen. https://www.intechopen.com/books/european-
23 24 537 25	local-pig-breeds-diversity-and-performance-a-study-of-project-treasure
<sup>26</sup> 538 27	Ceballos F.C., Joshi P.K., Clark D.W., Ramsay M. & Wilson J.F. (2018) Runs of homozygosity:
28 29 539	Windows into population history and trait architecture. Nature Review Genetics 19, 220–34.
30 31 540 32	Chang C.C., Chow C.C., Tellier L.C.A.M., Vattikuti S., Purcell S.M. & Lee J.J. (2015) Second-
33 541 34	generation PLINK: Rising to the challenge of larger and richer datasets. <i>Gigascience</i> 4, 7.
<sup>35</sup> 542 36	Charlesworth D. & Willis J.H. (2009) The genetics of inbreeding depression. Nature Review Genetics
37 38 543	10, 793–6.
39 40 544 41	Chen E.Y., Tan C.M., Kou Y., Duan Q., Wang Z., Meirelles G.V., Clark N.R. & Ma'ayan A. (2013)
<sup>42</sup> 545 43	Enrichr: interactive and collaborative HTML5 gene list enrichment analysis tool. BMC
44 45 546	Bioinformatics 14, 1–14.
46 47 547 48	Chen C., Liu C., Xiong X., Fang S., Yang H., Zhang Z., Ren J., Guo Y. & Huang L. (2018) Copy
49 548 50	number variation in the MSRB3 gene enlarges porcine ear size through a mechanism involving
51 52 549	miR-584-5p. Genetics Selection Evolution 50, 72.
53 54 550 55	Fabuel E., Barragán C., Silió L., Rodríguez M.C. & Toro M.A. (2004) Analysis of genetic diversity
55 56 551 57	and conservation priorities in Iberian pigs based on microsatellite markers. Heredity 93, 104-
<sup>58</sup> 552 59 60	13.

Page 23 of 36

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### **Animal Genetics**

3 553 4	Ferenčaković M., Sölkner J. & Curik I. (2013) Estimating autozygosity from high-throughput
5 6 554	information: Effects of SNP density and genotyping errors. <i>Genetics Selection Evolution</i> 45, 42.
7 8 555 9	Fontanesi L. & Russo V. (2013) Molecular genetics of coat colour in pigs. Acta Agriculturae
10 556 11	Slovenica Suppl. 4, 15–20.
<sup>12</sup> 557 13	Fontanesi L., Galimberti G., Calò D.G., Fronza R., Martelli P.L., Scotti E., Colombo M., Schiavo G.,
14 15 558 16	Casadio R., Buttazzoni L. & Russo V. (2012) Identification and association analysis of several
17 559 18	hundred single nucleotide polymorphisms within candidate genes for back fat thickness in Italian
<sup>19</sup> 560 20	large white pigs using a selective genotyping approach. <i>Journal of Animal Science</i> <b>90</b> , 2450–64.
21 22 561 23	Fontanesi L., Schiavo G., Galimberti G., Calò D.G. & Russo V. (2014) A genomewide association
23 24 562 25	study for average daily gain in Italian large white pigs. <i>Journal of Animal Science</i> <b>92</b> , 1385–94.
26 563 27	Gibson J., Morton N.E. & Collins A. (2006) Extended tracts of homozygosity in outbred human
28 29 564 30	populations. Human Molecular Genetics 15, 789–95.
30 31 565 32	Gomez-Raya L., Priest K., Rauw W.M., Okomo-Adhiambo M., Thain D., Bruce B., Rink A., Torell
33 566 34	R., Grellman L., Narayanan R. & Beattie C.W. (2008) The value of DNA paternity identification
<sup>35</sup> 567 36	in beef cattle: Examples from Nevada's free-range ranches. Journal of Animal Science 86, 17-
37 38 568 39	24.
	Gorssen W., Meyermans R., Buys N. & Janssens S. (2020) SNP genotypes reveal breed substructure,
42 570 43	selection signatures and highly inbred regions in Piétrain pigs. Animal Genetics 51, 32–42.
44 45 571	Grilz-Seger G., Druml T., Neuditschko M., Dobretsberger M., Horna M. & Brem G. (2019) High-
46 47 572 48	resolution population structure and runs of homozygosity reveal the genetic architecture of
49 573 50	complex traits in the Lipizzan horse. BMC Genomics 20, 174.
51 52 574	Grilz-Seger G., Druml T., Neuditschko M., Mesarič M., Cotman M. & Brem G. (2019) Analysis of
53 54 575 55	ROH patterns in the Noriker horse breed reveals signatures of selection for coat color and body
56 576 57 58 59 60	size. Animal Genetics 50, 334–46.

2 3 4	577	Grilz-Seger G., Mesarič M., Cotman M., Neuditschko M., Druml T. & Brem G. (2018) Runs of
5 6 7	578	homozygosity and population history of three horse breeds with small population size. Journal
	579	of Equine Veterinary Science 71, 27–34.
	580	Kios D., van Marle-Köster E. & Visser C. (2012) Application of DNA markers in parentage
	581	verification of Boran cattle in Kenya. Tropical Animals and Health Prodution 44, 471-6.
	582	Kirin M., McQuillan R., Franklin C.S., Campbell H., Mckeigue P.M. & Wilson J.F. (2010) Genomic
16 17 18	583	runs of homozygosity record population history and consanguinity. PLoS One 5, e13996.
	584	Kristensen T.N., Pedersen K.S., Vermeulen C.J. & Loeschcke V. (2010) Research on inbreeding in
	585	the "omic" era. Trends in Ecology and Evolution 25, 44-52.
23 24 25	586	Ligges U. & Mächler M. (2003) Scatterplot3d - An R package for visualizing multivariate data.
	587	Journal of Statistical Software 8, 1–20.
28 29	588	Marras G., Gaspa G., Sorbolini S., Dimauro C., Ajmone-Marsan P., Valentini A., Williams J.L. &
	589	Macciotta N.P.P. (2015) Analysis of runs of homozygosity and their relationship with inbreeding
32 33 34	590	in five cattle breeds farmed in Italy. Animal Genetics 46, 110–21.
	591	Mastrangelo S., Sardina M.T., Tolone M., Di Gerlando R., Sutera A.M., Fontanesi L. & Portolano B.
	592	(2018) Genome-wide identification of runs of homozygosity islands and associated genes in
39 40 41	593	local dairy cattle breeds. Animal 12, 2480–8.
	594	Meyermans R., Gorssen W., Buys N. & Janssens S. (2020) How to study runs of homozygosity using
	595	plink? A guide for analyzing medium density snp data in livestock and pet species. BMC
	596	Genomics 21, 94.
48 49 50	597	Mikawa S., Morozumi T., Shimanuki S.I., Hayashi T., Uenishi H., Domukai M., Okumura N. &
	598	Awata T. (2007) Fine mapping of a swine quantitative trait locus for number of vertebrae and
	599	analysis of an orphan nuclear receptor, germ cell nuclear factor (NR6A1). Genome Research 17,
55 56 57 58	600	586–93.
59 60		

1 2		
3 4	601	Mikawa S., Sato S., Nii M., Morozumi T., Yoshioka G., Imaeda N., Yamaguchi T., Hayashi T. &
5 6	602	Awata T. (2011) Identification of a second gene associated with variation in vertebral number
7 8 9	603	in domestic pigs. BMC Genetics 12, 5.
-	604	Muñoz M., Bozzi R., García F., Núñez Y., Geraci C., Crovetti A., García-Casco J., Alves E., Škrlep
12 13	605	M., Charneca R., Martins J.M., Quintanilla R., Tibau J., Kušec G., Djurkin-Kušec I., Mercat
	606	M.J., Riquet J., Estellé J., Zimmer C., Razmaite V., Araújo J.P., Radović Č., Savić R., Karolyi
16 17 18	607	D., Gallo M., Čandek-Potokar M., Fontanesi L., Fernández A.I. & Óvilo C. (2018) Diversity
	608	across major and candidate genes in European local pig breeds. PLoS One 13, e0207475.
	609	Muñoz M., Bozzi R., García-Casco J., Núñez Y., Ribani A., Franci O., García F., Škrlep M., Schiavo
23 24 25	610	G., Bovo S., Utzeri V.J., Charneca R., Martins J.M., Quintanilla R., Tibau J., Margeta V.,
	611	Djurkin-Kušec I., Mercat M.J., Riquet J., Estellé J., Zimmer C., Razmaite V., Araújo J.P.,
	612	Radović Č., Savić R., Karolyi D., Gallo M., Čandek-Potokar M., Fernández A.I., Fontanesi L.
30 31 32	613	& Óvilo C. (2019) Genomic diversity, linkage disequilibrium and selection signatures in
	614	European local pig breeds assessed with a high density SNP chip. Scientific Reports 9, 13546.
35 36	615	Peripolli E., Metzger J., De Lemos M.V.A., Stafuzza N.B., Kluska S., Olivieri B.F., Feitosa F.L.B.,
	616	Berton M.P., Lopes F.B., Munari D.P., Lôbo R.B., Magnabosco C.D.U., Di Croce F., Osterstock
39 40 41	617	J., Denise S., Pereira A.S.C. & Baldi F. (2018) Autozygosity islands and ROH patterns in Nellore
	618	lineages: Evidence of selection for functionally important traits BMC Genomics 19, 680.
	619	Peripolli E., Munari D.P., Silva M.V.G.B., Lima A.L.F., Irgang R. & Baldi F. (2017) Runs of
46 47 48	620	homozygosity: current knowledge and applications in livestock. Animal Genetics 48, 255-71
	621	Porter V. (1993) Pigs: A handbook to the breeds of the world. Comstock Publishing Associates,
	622	Cornell University Press, Ithaca, New York, 256 pp., ISBN 0-8014-2920-X.
53 54 55	623	Purfield D.C., McParland S., Wall E. & Berry D.P. (2017) The distribution of runs of homozygosity
	624	and selection signatures in six commercial meat sheep breeds. PLoS One 12, e0176780.
	625	Rubin C.J., Megens H.J., Barrio A.M., Maqbool K., Sayyab S., Schwochow D., Wang C., Carlborg
60	626	Ö., Jern P., Jørgensen C.B., Archibald A.L., Fredholm M., Groenen M.A.M. & Andersson L.

2		
4	627	(2012) Strong signatures of selection in the domestic pig genome. Proceedings of the National
0	628	Academy of Sciences of the United States of America 109, 19529–36.
7 8 9	629	Schiavo G., Bovo S., Bertolini F., Tinarelli S., Dall'Olio S., Nanni Costa L., Gallo M. & Fontanesi
	630	L. (2020a) Comparative evaluation of genomic inbreeding parameters in seven commercial and
13	631	autochthonous pig breeds. Animal 14, 915–20.
	632	Schiavo G., Bovo S., Bertolini F., Dall'Olio S., Nanni Costa L., Tinarelli S., Gallo M. & Fontanesi
16 17 18	633	L. (2020b) Runs of homozygosity islands in Italian cosmopolitan and autochthonous pig breeds
	634	identify selection signatures in the porcine genome. <i>Livestock Science</i> <b>240</b> , 104219.
	635	Silió L., Rodríguez M.C., Fernández A., Barragán C., Benítez R., Óvilo C. & Fernández A.I. (2013)
23 24 25	636	Measuring inbreeding and inbreeding depression on pig growth from pedigree or SNP-derived
	637	metrics. Journal of Animal Breeding and Genetics 130, 349–60.
	638	Szmatoła T., Gurgul A., Ropka-Molik K., Jasielczuk I., Zabek T. & Bugno-Poniewierska M. (2016)
30		
	639	Characteristics of runs of homozygosity in selected cattle breeds maintained in Poland. <i>Livestock</i>
32	639 640	Science 188, 72–80.
32 33 34 35 36	640	
32 33 34 35 36 37 38	640	<i>Science</i> <b>188</b> , 72–80.
32 33 34 35 36 37 38 39 40 41	640 641 642 643	Science 188, 72–80. VanRaden P.M., Olson K.M., Null D.J. & Hutchison J.L. (2011) Harmful recessive effects on fertility
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<ul> <li>32</li> <li>33</li> <li>34</li> <li>35</li> <li>36</li> <li>37</li> <li>38</li> <li>39</li> <li>40</li> <li>41</li> <li>42</li> <li>43</li> <li>44</li> <li>45</li> </ul>	640 641 642 643	<ul> <li>Science 188, 72–80.</li> <li>VanRaden P.M., Olson K.M., Null D.J. &amp; Hutchison J.L. (2011) Harmful recessive effects on fertility detected by absence of homozygous haplotypes. <i>Journal of Dairy Science</i> 94, 6153–61.</li> <li>VanRaden P.M. (2008) Efficient methods to compute genomic predictions. <i>Journal of Dairy Science</i></li> </ul>
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32 33 34 35 36 37 38 39 40 41 42 43 44 45 46 47 48 49 50 51 52 53 54	640 641 642 643 644 645 646 647	<ul> <li>Science 188, 72–80.</li> <li>VanRaden P.M., Olson K.M., Null D.J. &amp; Hutchison J.L. (2011) Harmful recessive effects on fertility detected by absence of homozygous haplotypes. <i>Journal of Dairy Science</i> 94, 6153–61.</li> <li>VanRaden P.M. (2008) Efficient methods to compute genomic predictions. <i>Journal of Dairy Science</i> 91, 4414–23.</li> <li>Wei T. &amp; Simko V. (2017) R package "corrplot": Visualization of a Correlation Matrix (Version 0.84). Available from https://github.com/taiyun/corrplot.</li> <li>Wright S. (1922) Coefficients of Inbreeding and Relationship. <i>American Naturalist</i> 56, 330–8.</li> </ul>
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 **Tables** 

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<b>Table 1.</b> Runs of homozygosity (ROH) parameters calculated in the 23 pig breeds obtained without
any pruning for minor allele frequency (MAF), i.e. MAF ≥0.00. Parameters calculated using MAF

>0.01 are reported in Table S4.

Breed	Acronym	nROH (SD) <sup>1</sup>	$L_{ROH} (SD)^2$	S <sub>ROH</sub> (SD) <sup>3</sup>
Alentejana	AL	50.90 (10.67)	6.49 (2.48)	339.97 (167.31)
Apulo-Calabrese	AC	56.74 (11.67)	14.21 (3.60)	813.75 (266.55)
Basque	BA	106.62 (9.36)	7.21 (1.13)	764.56 (105.38)
Bísara	BI	43.88 (12.93)	7.59 (2.67)	352.18 (211.11)
Black Slavonian	BS	36.61 (14.72)	8.75 (3.29)	336.98 (230.97)
Casertana	CA	45.34 (11.20)	12.63 (4.04)	595.06 (268.90)
CintaSenese	CS	55.62 (15.47)	7.75 (2.28)	424.32 (144.99)
Gascon	GA	75.08 (8.52)	6.97 (1.06)	522.14 (89.18)
Iberian	IB	51.38 (11.97)	6.50 (2.25)	341.52 (148.95)
Krškopolje	KR	34.96 (7.36)	8.62 (2.72)	306.47 (138.31)
Lithuanian indigenous wattle	LIW	42.69 (7.07)	7.69 (1.74)	330.44 (98.97)
Lithuanian White Old Type	LWOT	56.27 (10.16)	6.59 (1.82)	373.55 (133.34)
Majorcan Black	MB	48.50 (10.47)	6.58 (1.95)	327.89 (147.08)
Mora Romagnola	MR	70.35 (7.37)	14.38 (2.48)	1003.13 (139.75
Moravka	МО	30.14 (12.34)	8.48 (4.36)	289.36 (220.73)
Nero Siciliano	NS	24.15 (10.00)	7.30 (4.91)	207.33 (208.19)
Sarda	SA	27.46 (10.26)	7.77 (4.70)	246.77 (221.24)
Schwäbisch-Hällisches	SHS	49.14 (6.63)	7.28 (2.13)	360.16 (123.64)
Swallow-Bellied Mangalitsa	SBMA	49.96 (8.11)	9.75 (2.04)	483.27 (115.50)
Turopolje	TU	79.76 (15.31)	11.91 (1.78)	955.04 (242.37)
Italian Duroc	IDU	104.00 (10.49)	6.33 (1.03)	655.35 (106.75)
Italian Landrace	ILA	65.56 (8.86)	5.27 (1.08)	347.80 (92.75)
Italian Large White	ILW	62.46 (12.90)	5.52 (1.00)	349.22 (107.11)

<sup>1</sup> nROH: the average total number of ROH and the standard deviation (SD) calculated for each breed. <sup>2</sup>  $L_{ROH}$ : the average length of ROH (in Mb) considering all length classes and the standard deviation

(SD) calculated for each breed.

 $^{3}$  S<sub>ROH</sub>: the average sum of all ROH segments (in Mb) by animals considering all length classes and

the standard deviation (SD) calculated for each breed.

		tion is in parenthes	51S.	
Breed	F <sub>ROH1</sub>	F <sub>ROH4</sub>	F <sub>ROH8</sub>	F <sub>ROH1</sub>
Alentejana	0.139 (0.072)	0.110 (0.071)	0.084 (0.062)	0.059
Apulo-Calabrese	0.332 (0.111)	0.314 (0.110)	0.281 (0.102)	0.229
Basque	0.312 (0.042)	0.261 (0.052)	0.194 (0.053)	0.120
Bísara	0.144 (0.093)	0.122 (0.082)	0.098 (0.081)	0.071
Black Slavonian	0.138 (0.091)	0.121 (0.091)	0.101 (0.092)	0.072
Casertana	0.243 (0.112)	0.226 (0.110)	0.202 (0.110)	0.162
Cinta Senese	0.173 (0.064)	0.147 (0.063)	0.111 (0.052)	0.075
Gascon	0.213 (0.042)	0.175 (0.042)	0.132 (0.041)	0.087
Iberian	0.139 (0.063)	0.111 (0.061)	0.082 (0.060)	0.056
Krškopolje	0.125 (0.061)	0.109 (0.060)	0.089 (0.063)	0.065
Lithuanian indigenous wattle	0.135 (0.042)	0.114 (0.040)	0.089 (0.044)	0.060
Lithuanian White Old Type	0.152 (0.052)	0.122 (0.050)	0.093 (0.051)	0.063
Majorcan Black	0.134 (0.061)	0.108 (0.060)	0.081 (0.051)	0.055
Mora Romagnola	0.409 (0.062)	0.386 (0.062)	0.345 (0.060)	0.286
Moravka	0.118 (0.092)	0.103 (0.091)	0.087 (0.080)	0.068
Nero Siciliano	0.085 (0.084)	0.073 (0.082)	0.059 (0.081)	0.043
Sarda	0.101 (0.092)	0.088 (0.094)	0.073 (0.092)	0.053
Schwäbisch-Hällisches	0.147 (0.051)	0.120 (0.052)	0.093 (0.052)	0.065
Swallow-Bellied Mangalitsa	0.197 (0.052)	0.175 (0.050)	0.146 (0.050)	0.107
Turopolje	0.390 (0.101)	0.362 (0.101)	0.311 (0.093)	0.238
Italian Duroc	0.267 (0.043)	0.211 (0.041)	0.157 (0.041)	0.104
Italian Landrace	0.142 (0.042)	0.104 (0.040)	0.069 (0.031)	0.041
Italian Large White	0.143 (0.041)	0.106 (0.042)	0.075 (0.040)	0.046

Average length

 $(Mb)^3$ 

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2.99 (2.25)

2.37 (1.84)

1.90 (1.36)

0.88(0.44)

1.46 (1.52)

2.67 (2.42)

2.33 (2.00)

2.45 (1.49)

2.33 (2.14)

2.79 (2.00)

2.04 (2.19)

2.27 (1.87)

3.09 (3.41)

2.12 (2.65)

1.85 (1.83)

2.14 (1.76)

2.93 (1.89)

4.75 (3.50)

4.89 (6.48)

2.34 (2.48)

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<sup>3</sup> 669 4	Table 3. The number of runs of				
5 6 670 7	by ROH islands identified in the				
7 8 9	Breed				
10	Alentejana				
11	Apulo-Calabrese				
12	Basque				
13	Bísara				
14	Black Slavonian				
15 16					
17	Casertana				
18	Cinta Senese				
19	Gascon				
20	Iberian				
21	Krškopolje				
22	Lithuanian indigenous				
23	wattle				
24	Lithuanian White Old Type				
25 26	Majorcan Black				
20	Mora Romagnola				
28	Moravka				
29	Nero Siciliano				
30	Sarda				
31	Schwäbisch-Hällisches				
32					
33	Swallow-Bellied Mangalitsa				
34 35	Turopolje				
36	Italian Duroc				
37	Italian Landrace				
38	Italian Large White				
39 671 40	<sup>1</sup> Frequency of the SNPs in a R				
41 672 42	has been calculated dividing the				
43 44 673	of animals retained after genot				
45 46 674 47	<sup>2</sup> Sum of the length of the chro				
48 675 49	<sup>3</sup> Average length of the ROH i				
<sup>50</sup> 676 51	The three blocs indicate the tw				
52 53 677	block, there is information abo				
54 55 678 56	the number of islands Identified				
57 679 58	length of islands.				

1 2

<sup>3</sup> 669 Table 3. The number of runs of homozygosity (ROH) islands and information on the genome covered

570	by ROH islands i	dentified in the 23 pig	breeds with the method	d that used the S <sub>ROH</sub> based	-threshold.
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N. of ROH

islands

12

0

7

7

3

7

26

12

15

15

15

22

12

4

9

4

0

17

8

17

19

32

Genome

35.88

16.58

13.32

2.64

10.23

69.37

27.99

36.74

34.89

41.81

44.84

27.23

12.34

19.11

7.41

36.40

23.41

80.82

92.85

75.03

-

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covered (Mb)<sup>2</sup>

Frequency<sup>1</sup>

19/48 (40%)

38/53 (72%)

36/39 (92%)

20/48 (42%)

19/49 (39%)

29/53 (55%)

23/53 (43%)

27/48 (56%)

19/48 (40%)

18/52 (35%)

19/48 (40%)

21/48 (44%)

19/48 (40%)

46/48 (96%)

17/49 (35%)

14/48 (29%)

16/48 (33%)

20/49 (41%)

25/50 (50%)

44/50 (88%)

32/48 (67%)

20/48 (42%)

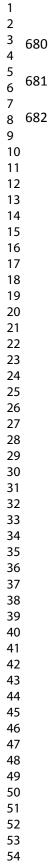
alian Large White 20/48 (42%) 12 46.51 3.88 (2.57) requency of the SNPs in a ROH that identifies the threshold to declare a ROH island. The frequency s been calculated dividing the number of animals needed to reach the define level by the number animals retained after genotyping (see Table S2). Sum of the length of the chromosome regions in the genome covered by ROH islands in Mb.

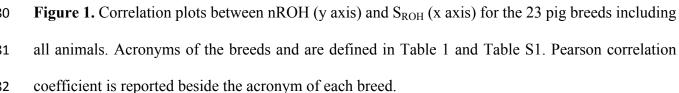
Average length of the ROH islands (standard deviation) in Mb.

the three blocs indicate the two different thresholds that can be used to define an island. For each

ock, there is information about: the number of animals that is used as threshold to define ad Island,

e number of islands Identified, the total length of genome that is covered by islands, the average





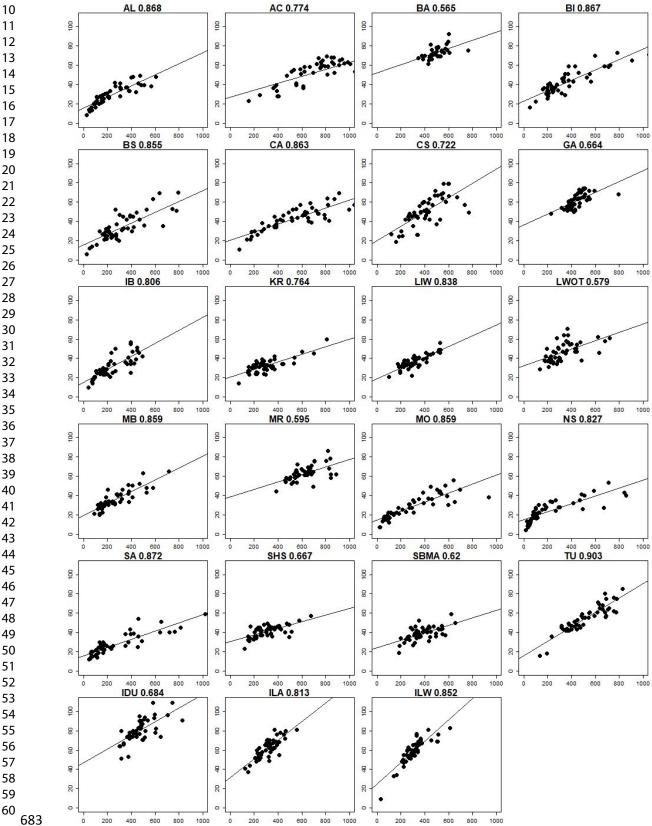


Figure 2. Proportion of runs of homozygosity of different class size in the 23 pig breeds. ROH classes
were defined according to their size: 1–2, 2–4, 4–8, 8–16 and >16 Mb, identified as ROH1–2, ROH2–
4, ROH4–8, ROH8–16 and ROH>16, respectively.

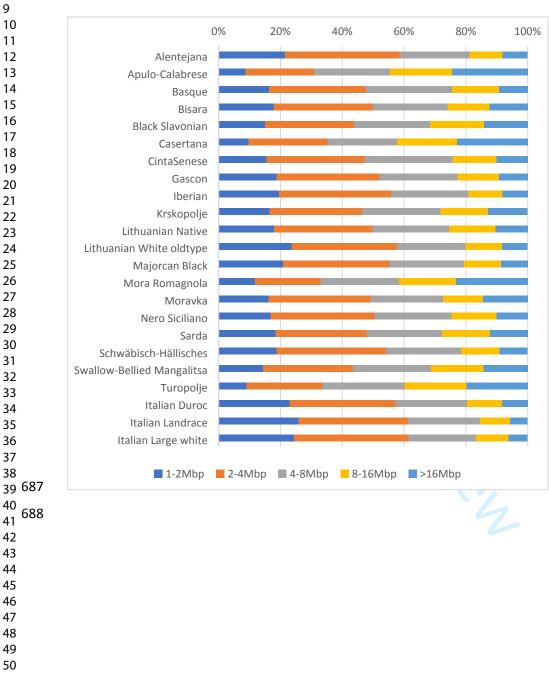
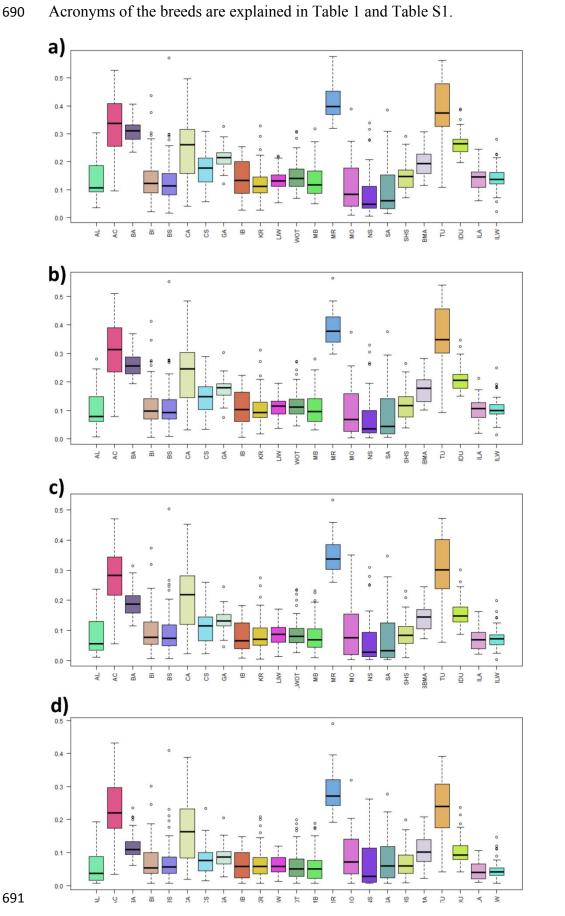
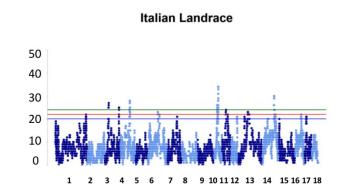


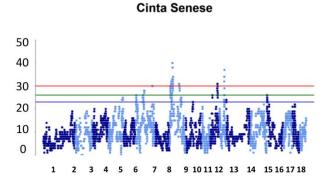
Figure 3. Boxplots of the F<sub>ROH</sub> distribution in the 23 breeds: a) F<sub>ROH1</sub>; b) F<sub>ROH4</sub>; c) F<sub>ROH8</sub>; d) F<sub>ROH16</sub>.

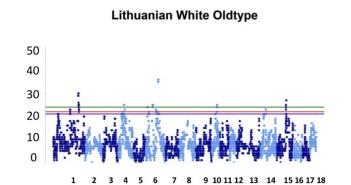


### Animal Genetics

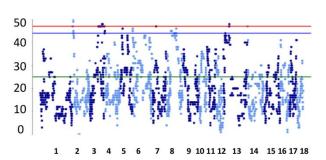
**Figure 4.** Manhattan plots showing ROH islands in a few analysed pig breeds with extreme patterns. The red line indicates the  $S_{ROH}$ -based threshold, the blue line indicates the frequency corresponding to the top 1% most frequent SNP in the population, the green line indicates the 50% of individuals within the population. The y axes indicate the number of animals carrying that SNP in a ROH.



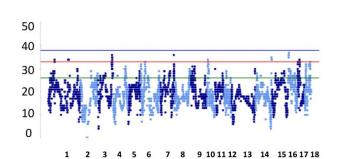




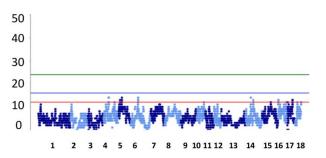


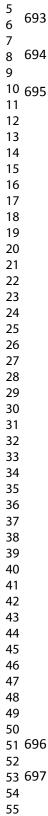


Apulo-Calabrese

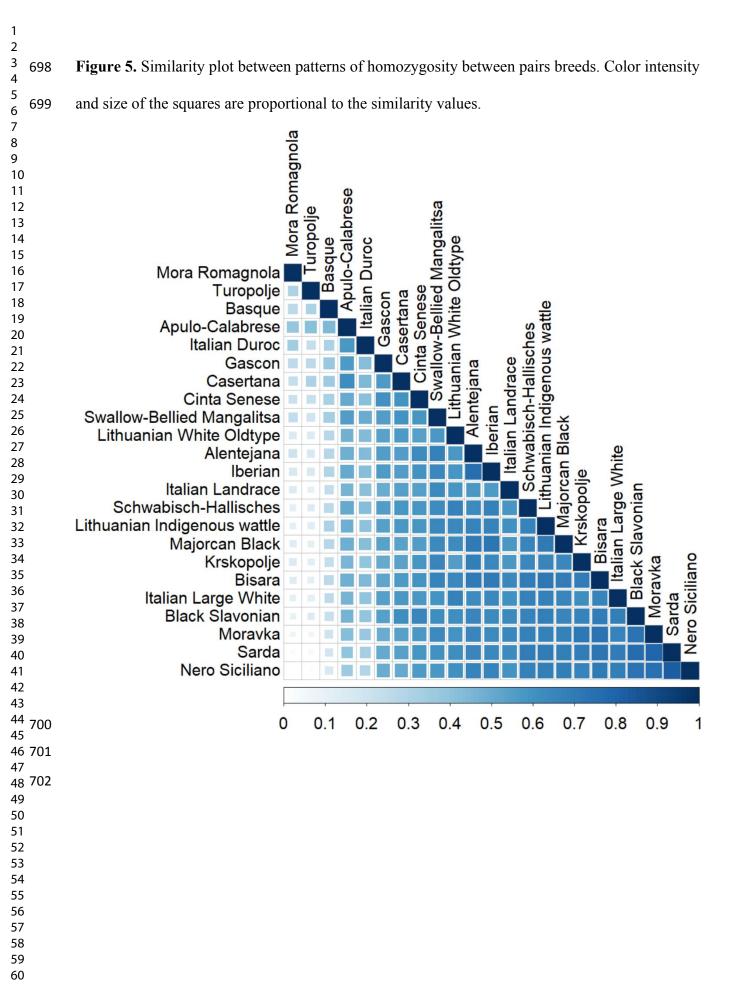








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# 3 Supporting information legend

Table S1: Analysed breeds, their country and region of origin and other information useful to describethe breeds.

**Table S2**: Number of animals and analysed SNP before and after the filtering steps.

**Table S3.** Effective population size (*Ne*) calculated for each breed.

- Table S4. Runs of homozygosity (ROH) parameters using minor allele frequency (MAF)  $\geq 0.01$
- 7 709 Table S5. Minimum and maximum values for the number and size of ROH (nROH and  $L_{ROH}$ ,

710 respectively) and for the sum of all ROH segments by animals.

**Table S6.** Proportion of the five different runs of homozygosity (ROH) classes for each breed.

4 712 **Table S7.** Mean  $F_{ROH}$  values calculated using different ROH lengths and MAF >0.01.

<sup>6</sup>713 **Table S8.** Average values for several genomic inbreeding measures.

<sup>8</sup> 714 **Table S9.** Correlation between all genomic inbreeding parameters in all breeds.

**Table S10.** The number ROH islands and information on the genome covered.

<sup>3</sup> 716 **Table S11**. ROH Islands and annotations (Excel file).

717 **Table S12.** Results of the gene enrichment analysis on all ROH Islands.

**Table S13.** Results of the gene enrichment analysis on ROH Islands that overlapped previous work

719 regions identifying selection signature.

**Figure S1**. Multidimensional scaling (MDS) plot of the 23 pig breeds.

Figure S2. Genomic inbreeding based on  $F_{ROH}$  across chromosomes ( $F_{ROHSS}$ ).

**Figure S3.** Boxplot of the Inbreeding Coefficients estimated with all the different methods.

**Figure S4.** Boxplot of the Inbreeding Coefficients estimated with all the different methods.

**Figure S5.** Manhattan plots showing ROH island patterns in all investigated pig breeds. The red line

indicates the  $S_{ROH}$ -based threshold, the blue line indicates the frequency corresponding to the top 1%

most frequent SNP in the population, the green line indicates the 50% of individuals within the

population. The y axes indicate the number of animals carrying that SNP in a ROH.

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