Whole blood transcriptome analysis reveals footprints of cattle adaptation to sub-arctic

conditions

Kisun Pokharel[1], Melak Weldenegodguad[1,2], Ruslan Popov[3], Mervi Honkatukia[1,4], Hanna Huuki[1],

Heli Lindeberg[5], Jaana Peippo[1], Tiina Reilas[1], Stepan Zarovnyaev[6], and Juha Kantanen[1*]

[1]Production Systems, Natural Resources Institute Finland (Luke), Myllytie 1, FI-31600, Jokioinen,
Finland

[2]Department of Environmental and Biological Sciences, University of Eastern Finland, P.O. Box
1627, FI-70311 Kuopio, Finland

[3]Yakutian Research Institute of Agriculture (FGBNU Yakutskij NIISH), 67001 Yakutsk, ul.
Bestyzhevo-Marlinskogo 23/1, Russia, The Sakha Republic (Yakutia)

[4]The Nordic Genetic Resources Center (Nordgen), P.O. Box 115, NO-1431 Ås, Norway

[5]Production Systems, Natural Resources Institute Finland (Luke), Halolantie 31A, FI-71750,
Maaninka, Finland

[6]GBU Saha Agroplem, 67700 Yakutsk, ul. Ordzhonkidze 20/204, Russia, The Sakha Republic
(Yakutia)

[*]Corresponding author

Juha Kantanen

Phone: +358 (0)29 5326210

Email: juha.kantanen@luke.fi

Summary

Indigenous cattle breeds in northern Eurasia have adapted to harsh climate conditions. The local

breeds are important genetic resources with cultural and historical heritages, and therefore, their

preservation and genetic characterization are important. In this study, we aim to identify genes and

biological processes that are important for their adaptation to the cold and harsh conditions. For this

purpose, we profiled the whole-blood transcriptome of two native breeds and one commercial breed

using high-throughput RNA sequencing. More than 15,000 genes were identified, of which 2, 89 and

162 genes were significantly upregulated exclusively in Northern Finncattle, Yakutian cattle and

Holstein cattle, respectively. The functional classification of these significantly differentially

expressed genes identified several biological processes and pathways related to signalling

mechanisms, cell differentiation, and host-pathogen interactions that, in general, point towards

immunity and disease resistance mechanisms. The gene expression pattern observed in Northern

Finncattle was more similar to that of Yakutian cattle, despite sharing similar living conditions as the

Holstein cattle included in our study. In conclusion, our study identified unique biological processes

in these breeds that help them to adapt and survive in sub-arctic environments.

Keywords: adaptation, gene expression, Yakutian cattle, Northern Finncattle, Holstein cattle, blood

Running title: Genomics of cattle adaptation to sub-arctic conditions

Introduction

As a result of natural and human-led selection, domestic animals have been able to adapt, survive, be

productive and reproduce in challenging environments (Mirkena et al., 2010). Animals in the northern

hemisphere (Lapland, northern Russia and Siberia) are used for food production and other socio-

cultural needs and, thus, have been the basis of human life in the North. These animals may have

different biological capacities to adapt to extremes in temperature, daylight and feed availability.

However, there has been growing interest towards improving local breeds by introducing genetic

material from superior breeds, with a preference towards commercial breeds. Intensive breeding and

the replacement of native breeds with commercial breeds may appear to be advantageous at first, but

these practices will have long-lasting consequences. In addition to irreversibly losing the unique

genetic resources of native breeds, we will also be losing the cultural and historical heritage

associated with those local breeds. The signatures of adaptation, as well as the history of formation,

that are encoded in the native breeds may also fade away (Gaouar et al., 2015). Therefore, the local

breeds must be genetically characterized and preserved.

Here, we have studied two native cattle breeds (Northern Finncattle and Yakutian cattle) and one

international breed (Holstein cattle). All of these breeds are used for milk and meat production.

Previous studies using genetic markers have indicated the genetic distinctiveness among these breeds

(Li et al., 2007; Li and Kantanen, 2010). Among the three breeds, Holstein cattle have high economic

importance, due to high productivity, and have the shortest adaptation history. Holstein cattle are the

most popular dairy breed globally, with an intensive selection programme and high milk production;

although they originated in a temperate climate, they have adapted to different parts of the world and

can survive in varying climatic conditions. Northern Finncattle and Yakutian cattle have economic,

social and cultural values (Kantanen et al., 2015). The Northern Finncattle breed is native to Northern

Finland and Finnish Lapland. The breed nearly became extinct during the1970s but is currently

maintained (current census is 850 cows) through active *in vivo* and *in vitro* conservation activity. The

Yakutian cattle are characterized by being purebred aboriginal native cattle. Adult Yakutian cows

(current population is approx. 1,000 animals) typically weigh 350-400 kg, and their height is 111 cm,

on average. The animals are well adapted to harsh Siberian conditions, where the temperature falls

below -50°C in long winters (Kantanen et al., 2009).

High-throughput RNA sequencing (RNA-Seq) has been proven to be an efficient method for studying

the regulation of gene expression (Wang et al., 2009). In recent years, RNA-Seq has been applied to

domestic animals, with a major focus on productivity traits (Bai et al., 2016; Li et al., 2016; Pokharel

et al., 2018; Silva-Vignato et al., 2017). A number of transcriptome studies have been conducted in

different Holstein breeds (Bai et al., 2016; Li et al., 2016; Sandri et al., 2015), but to date, there are no

reports on gene expression studies in either Yakutian cattle or Northern Finncattle. In this study, we

have applied RNA-Seq technology and profiled the whole blood transcriptome of the three

aforementioned breeds to characterize their genetic differences.

Materials and methods

Sample collection

Animal handling procedures and sample collection were conducted in accordance with legal

regulations that were approved by the Russian authorization board (FS/UVN-03/163733/07.04.2016)

and the Animal Experiment Board in Finland (ESAVI/7034/04.10.07.2015). A 2.5-ml sample of

blood from each of three Yakutian cattle, three Northern Finncattle and three Holstein cattle cows was

collected into a PAXgene® Blood RNA IVD tube (Ref. PreAnalytiX®, Hombrechtikon, Switzerland)

during the winter period of 2016 and 2017 and stored at -18°C. The Finnish samples were taken in a

slaughterhouse at exsanguination, and the Yakutian samples were collected by jugular venipuncture in

the northern Yakutian village Sakkyryr (YC6) and central Yakutia Magaras (YC10 and YC11). All of

the animals included in this study were 4- to 8-year-old females, except for one 14-year-old Holstein

cow (HC3).

RNA extraction

The RNA was extracted with the PAXgene® Blood RNA kit (Ref. 762174, Ref. PreAnalytiX®,

Hombrechtikon, Switzerland), according to the kit manual, with minor adjustments to the protocol:

the samples were thawed at room temperature overnight, the initial centrifugation time at $5,000 \times g$

was increased to 15 minutes, the initial pellet was resuspended into double the amount of BR1 buffer

and divided into two separate reactions, the proteinase K incubation time was extended to 1 hour, and

the columns were incubated at room temperature in elution buffer for 10 minutes prior to elution. The

concentration and quality of the RNA was measured with a spectrophotometer (NanoDrop ND-1000,

Thermo Fisher Scientific, Delaware, USA), and the integrity of the RNA was measured with an

Agilent Bioanalyzer 2100 (Agilent, Waldbronn, Germany) using the Agilent 6000 RNA Nano kit

(Ref. 5067-1511, Agilent, Waldbronn, Germany). All samples selected for sequencing had an RNA

Integrity Number (RIN) value of at least 7.

Library preparation and sequencing

Library preparation and sequencing tasks were outsourced to the Finnish Functional Genomics Center

(FFGC) in Turku, Finland. The library preparation was performed according to Illumina's Truseq®

mRNA sample preparation guide protocol. The high quality of the libraries was confirmed with an

Advanced Analytical Fragment Analyzer, and the concentrations of the libraries were quantified using

Qubit® Fluorometric Quantitation, Life Technologies. The average RNA-Seq library fragments were

in the range of 250-350 bp. Only the good libraries with RNA Quality Number (RQN) value greater

than 7 were sequenced with a paired-end strategy to generate 75 bp reads.

Computational methods

Raw sequence reads were pre-processed using FastQC (Simon Andrews, n.d.) to determine the quality

of the data and to obtain an overview of the sequencing data. The outputs from FastQC were

summarized using Multiqc (Ewels et al., 2016). Processed reads were mapped against the latest

versions of the cattle reference genome UMD3.1 and transcriptome (Ensembl release 93) using STAR

v2.6 (Dobin et al., 2013). First, the genome indices were prepared, and mapping was performed with

default parameters. Moreover, a gene-level counts file for each sample was generated as part of the

star-alignment pipeline. Read counts were then processed using DESeq2 (Love et al., 2014) for gene

expression analysis, and all genes that had less than 5 transcript counts were discarded. To identify

significantly differentially expressed genes between breed groups, we set an adjusted p-value of 0.05.

For the list of significantly differentially expressed genes, we retrieved additional information, such as

gene description and chromosomal location, using the Biomart Bioconductor package (Durinck et al.,

2009; Smedley et al., 2009). Finally, the significantly differentially expressed genes were assessed for

their functional roles. Gene ontology (GO) terms and the Kyoto Encyclopaedia of Genes and

Genomes (KEGG) pathway analyses were performed using the ClueGO (Bindea et al., 2009) plugin

in Cytoscape (v3.6.1) (Shannon et al., 2003). In addition to the default parameters in ClueGO, we

chose to observe GO terms in the range of level 3 and 5. Similarly, for the GO term or KEGG

pathway to be displayed, at least three genes and a minimum of 4% of the total genes should be

present in our list. In addition, the terms or pathways shared by 50% of the genes from our list were

grouped together using Kappa statistics (Kappa score threshold of 0.4).

Results and discussion

From nine samples, we obtained 32.2 Gb of RNA-Seq data. More than 87% of the reads from each

sample mapped to the cattle reference genome, with Holstein samples having more than 91% mapping

rates. The slightly higher mapping rate for Holstein samples could be because that breed is closer to

the Hereford breed, which is the source of the reference assembly (Zimin et al., 2009). With

sequencing costs becoming cheaper, having breed-specific reference genomes would be useful in the

future.

A total of 15,960 genes (Table S1) were expressed in our data, which covers 64.8% of known

(24,616) cattle genes reported in the annotation file. At the individual level, we observed that both the

highest (FC8) and the lowest (FC7) number of genes were expressed in Northern Finncattle samples

(Table 1). We believe that a comparatively lower number of reads in the sample FC7 led to a lower

number of expressed genes. We did not observe any globin genes among the top most expressed

genes. Genes such as MHC class I heavy chain (BOLA), uncoupling protein 2 (UCP2), eukaryotic

translation elongation factor 3 (EEF2) and vimentin (VIM) were among the top most expressed genes.

These results confirmed that RNA extraction using PAXgene® worked well.

We assessed the co-expression status of the top 500 most expressed genes (i.e., the genes with highest

base mean values) in each of the three breed groups. As seen in the Venn diagram (Fig. 1), 386 of the

top expressed genes were present in all three breeds. Holstein samples had the highest number (n =

65) of unique genes, followed by Yakutian cattle (n = 56) and Northern Finncattle (n = 17). Moreover,

the results showed that Northern Finncattle are more closely related to Yakutian cattle (53 uniquely

shared genes) than to Holstein cattle (44 uniquely shared genes). In contrast, Yakutian cattle and

Holstein cattle shared the least number of unique genes.

We then looked more closely into the genes that were differentially expressed between any two of the

three breed groups and assessed what types of biological processes and/or biological pathways those

genes were associated with.

Differential gene expression between Yakutian cattle and Northern Finncattle

A total of 189 transcripts were significantly differentially expressed between Yakutian cattle and

Northern Finncattle (Table S2), of which 108 transcripts were upregulated in Yakutian cattle.

Transcripts of known genes, such as *C-C motif chemokine ligand 5* (*CCL5*), *perforin 1* (*PRF1*), *SLAM*

family member 7 (*SLAMF7*), *CD300 molecule-like family member b* (*CD300LB*), *sphingosine-1-*

phosphate receptor 5 (*S1PR5*) and *ATP binding cassette subfamily A member 13* (*ABCA13*), were

upregulated in Yakutian cattle. Genes upregulated in Northern Finncattle included *transglutaminase 3*

(*TGM3*), *5'-aminolevulinate synthase 2* (*ALAS2*), *ATP synthase inhibitory factor subunit 1* (*ATP5IF1*)

and *FK506 binding protein 5* (*FKBP5*). Moreover, a number of genes with unknown functions were

upregulated in Yakutian cattle (e.g., ENSBTAG00000047816, ENSBTAG00000038141,

ENSBTAG00000010300, and ENSBTAG00000046323) and in Northern Finncattle (e.g.,

ENSBTAG00000046981, ENSBTAG00000048145, ENSBTAG00000003754, and

ENSBTAG00000002450). Having a better annotation would certainly help in characterizing the

functions of these genes. The GO terms associated with the upregulated genes in Yakutian cattle

included "positive regulation of natural killer cell chemotaxis", "regulation of phagocytosis, negative

regulation of viral process", "negative regulation of cytokine production" and "granzyme-mediated

apoptotic signalling pathway" (Fig. 2a, Table S3). The KEGG pathways associated with upregulated

genes included "antigen processing and presentation", "natural killer cell mediated cytotoxicity" and

"graft-versus-host disease" (Table S4). Similarly, the GO terms associated with upregulated genes in

Northern Finncattle included "regulation of lipid transport", "erythrocyte differentiation", "regulation

of cytokine-mediated signalling pathway" and "regulation of protein maturation" (Fig. 2b, Table S5).

"Malaria" (associated genes *HBA*, *HBB* and *THBS1*) was the only KEGG pathway related to genes

upregulated in Northern Finncattle.

Differential gene expression between Yakutian cattle and Holstein cattle

The highest number of significantly differentially expressed genes was identified between Yakutian

cattle and Holstein cattle (Table S6). Given the phenotypic difference between the breeds, it was not

surprising to observe a large number of differentially expressed genes. Yakutian cattle have

phenotypic characteristics, such as a small and compact body size and a thick hair coat, which helps

them survive in a cold environment. In contrast, Holstein cattle have a bigger body size and a thin hair

coat and are thus not ready for a colder climate. In addition, Holstein cattle do not face a shortage of

fodder and are maintained inside the barn, whereas Yakutian cattle may suffer from food shortages

and are partly outside during the winter. The significant differences between Yakutian cattle and

Holstein cattle are in line with earlier reports (Decker et al., 2016; Yurchenko et al., 2018; Zinovieva

et al., 2016). Out of 1,418 genes that were significantly differentially expressed between the two

breeds, 594 genes were upregulated in Yakutian cattle. Because of the high number of differentially

expressed genes, we only considered those genes with a log2-fold change greater than or less than 1.5

for GO and KEGG pathway analyses. Genes that were upregulated in Yakutian cattle were associated

with GO terms such as "chemokine-mediated signalling pathway", "regulation of natural killer cell-

mediated immunity", "kinase activator activity" and "regulation of T cell activation" (Fig. 3a, Table

S7). In contrast, the genes that were upregulated in Holstein cattle were associated with GO terms

such as "inflammatory response", "regulation of neutrophil chemotaxis", "regulation of response to

external stimulus" and "secondary active transmembrane transporter activity" (Fig. 3c, Table S8). The

KEGG pathways associated with upregulated genes in Yakutian cattle included "natural killer cell-

mediated cytotoxicity", "African trypanosomiasis", "endocrine and other factor-regulated calcium

reabsorption", "cytosolic DNA-sensing pathway", "cell adhesion molecules (CAMs)" and "cytokine-

cytokine receptor interaction" (Fig. 3b, Table S9). Similarly, pathways such as "non-small cell lung

cancer", "legionellosis", "notch signalling pathway", "cytokine-cytokine receptor interaction",

"adipocytokine signalling pathway", "osteoclast differentiation", "protein digestion and absorption"

and "ABC transporters" were related to the genes that were upregulated in Holstein cattle (Fig. 3d,

Table S10).

Differential expression between Northern Finncattle and Holstein cattle

Between Finncattle and Holstein cattle, we observed 250 significantly differentially expressed

transcripts (Table S11), with majority the of genes (n = 180) being upregulated in Holstein cattle.

Only one GO term, "eicosanoid metabolic process" (associated genes *ALOX15*, *ALOX5*, *HPGD*), and

no KEGG pathways were identified for genes that were upregulated in Northern Finncattle. In

contrast, GO terms, such as "response to molecule of bacterial origin", "cellular response to

lipopolysaccharide", "negative regulation of haemopoiesis", and "response to mechanical stimulus"

(Fig. 4, Table S12), and KEGG pathways, such as "osteoclast differentiation", "complement and

coagulation cascades", "TNF signalling pathway" and "Toll-like receptor pathway", were associated

with upregulated genes in Holstein cattle (Table S13).

Uniquely differentially expressed genes

We noticed that there was overlap among many genes in more than one comparison (Fig. S1).

Therefore, we compiled a list of genes that were uniquely differentially expressed in each breed. The

number of uniquely differentially expressed genes (Table 2) in each breed followed a similar pattern

to that observed by the top 500 most expressed genes (Fig. 1).

Yakutian cattle had 139 uniquely differentially expressed genes, of which 89 were reported to be

upregulated. We noticed a number of cases where more than one gene from the same family was

upregulated. Such cases could provide higher confidence related to gene functions. The genes or

receptors with more than one member are as follows: chemokines (*CCL4*, *CCL5*), carbohydrate

sulfotransferases (*CHST1*, *CHST12*), chemokine receptors (*CX3CR1*, *CXCR6*), growth arrests (*GAS6*,

GAS7), granzymes (*GZMB* (two paralogs), *GZMM* and *GZMH*), insulin-like growth factor binding

proteins (*IGFBP4*, *IGFBP7*) and natural cytotoxicity triggering receptors (*NCR1*, *NCR3*). Among the

list were *BHLHE40* and *PRKCG*, which are linked to circadian rhythm. The importance of the

circadian clock is particularly applicable to Yakutian cattle. They must be metabolically prepared for

food scarcity during long winters (Ebling and Barrett, 2008).

Four granzyme transcripts and perforin were upregulated in Yakutian cattle. Granzymes are serine

proteases that are used by cytotoxic lymphocytes to destroy malignant and virus-infected cells.

Granzymes are transported into the cytoplasm of the target cell by *perforin 1* (*PRF1*), after which they

cleave specific proteins and trigger apoptosis (Johnson et al., 2003; MacDonald et al., 1999; Russell

and Ley, 2002). It has been suggested that the evolution of granzymes is related to species-specific

immune challenges (humans have 5 granzyme genes and mice have 10), which they maintain by gene

duplications and alterations in substrate specificity (Kaiserman et al., 2006). Three out of the six

granzymes known in cattle, (Yang et al., 2018) with two nearby paralogs of *granzyme B* (Fig. S2) and

PRF1, are all upregulated in Yakutian cattle. These findings suggest that Yakutian cattle have a very

solid granzyme-mediated immune system.

The number of upregulated genes was sufficient to identify significant GO terms. The GO terms

associated with the upregulated genes included "negative regulation of cytokine production",

"regulation of lymphocyte-mediated immunity", "positive regulation of leukocyte chemotaxis" and

"cellular response to interferon-gamma" (Fig. 5A, Table S14). Similarly, pathways such as "natural

killer cell-mediated cytotoxicity", "inflammatory bowel disease (IBD)", "African trypanosomiasis"

and "cytosolic DNA-sensing pathway" were related to uniquely upregulated genes in Yakutian cattle

(Fig. 5b, Table S15). Only one GO term, "regulation of lipid transport" (associated genes *ABCA1*,

ABCG1, *IRS2* and *THSB1*), and no KEGG pathways were found for the downregulated genes.

The majority (162/200) of the uniquely differentially expressed genes in Holstein cattle were found to

be upregulated. The list also includes a number of cases where more than one member of the gene

family was present. Genes encoding the solute carrier protein family (*SLC16A12*, *SLC16A6*,

SLC25A37, *SLC28A3*, *SLC2A9*, *SLC45A4*, *SLC6A6*, *SLC23A1*), ring finger proteins (*RNF149*,

RNF24), notch proteins (*NOTCH1*, *NOTCH2*, *NOTCH3*), lysine demethylases (*KDM4B*, *KDM6B*),

interferon receptors (*IFNAR1*, *IFNAR2*, *IFNGR1*), CD proteins (*CD101*, *CD164*, *CD300a*, *CD55*),

and others were upregulated in Holstein cattle samples. The GO terms associated with the upregulated

genes included "cellular response to lipopolysaccharide", "response to molecule of bacterial origin",

"negative regulation of leukocyte differentiation", "regulation of transcription from RNA polymerase

II promoter in response to stress and phosphatidylinositol binding" (Fig. 5C, Table S16). Similarly,

KEGG pathways such as "osteoclast differentiation", "Th1 and Th2 cell differentiation" and

"complement and coagulation cascades" were related to the upregulated genes (Fig. 5d, Table S17).

We did not find any GO terms or KEGG pathways associated with the downregulated genes.

Among other functions, membrane-bound solute carriers (SLCs) are key to maintaining physiological

processes, including nutrient uptake, waste removal and ion transport (Hediger et al., 2004). The

upregulated expression of eight SLC proteins in Holstein cattle samples suggest the highly important

function of these genes. A recent study showed that, during sloughing, amphibians increase the rate of

ion uptake to maintain internal homeostasis (Wu et al., 2017). We speculate that, with a comparatively

bigger body size and a thin hair coat, Holstein cattle may have developed a similar system for

maintaining internal homeostasis in response to lower temperatures. The upregulation of three notch

proteins (*NOTCH1*, *NOTCH2* and *NOTCH3*) and the biological processes associated with hypoxia

("cellular response to oxygen levels", "cellular response to hypoxia") is in line with earlier findings,

where the "notch signalling pathway" was activated in response to hypoxia (Hiyama et al., 2011).

This finding also suggests that Holstein cattle may not have adapted well to winter conditions, as

hypoxia is linked to low temperatures (Rocha and Branco, 1998). "Notch signalling pathway" is also

known to have important role in milk lactose metabolism, mammary gland development and lactation

which are all associated with high milk production traits of Holstein cattle (Politi et al., 2004; Yalcin-

Ozuysal et al., 2010). Moreover, *NOTCH1* and *NOTCH2* are part of the processes "Ras protein signal

transduction" and "regulation of ERK1 and ERK2 cascade".

Keratin 72 (*KRT72*, ENSBTAG00000007904) and *transmembrane protein 8 A* (*TMEM8A*,

ENSBTAG00000016588) were the only genes that were exclusively upregulated in Northern

Finncattle. *KRT72* is involved in hair formation, and the function of *TMEM8A* is unknown. Out of

seven downregulated genes, four IDs (ENSBTAG00000038233, ENSBTAG00000039691,

ENSBTAG00000046611, ENSBTAG00000000930) do not have gene descriptions available. The

remaining three genes are *acetylcholinesterase* (*ACHE*), *chemokine (C-C motif) ligand 3* (*CCL3*) and

Bos taurus regulator of G-protein signalling 2, 24 kDa (*RGS2*). The RGS family of proteins is

involved in fine-tuning the signalling activities of G-protein-coupled receptors (GPCRs) (Berman et

al., 1996; Blumer, 2004). An increased level of *RGS2* expression is linked to the regulation of

numerous biological activities, such as immune responses, bone formation, cardiovascular function

and anxiety (Zhang and Mende, 2014). *CCL3*, *CCL4* and *CCL5* are members of the CC chemokine

family of proteins, which exhibit proinflammatory activities. Moreover, *CCL3* inhibits the

proliferation of haematopoietic stem/progenitor cells (HSPCs) (Cook, 1996). *ACHE* has numerous

functions, and its role in muscle development, neuritogenesis, cell adhesion, the activation of

dopamine neurons, and amyloid fibre assembly has been reviewed earlier (Soreq and Seidman, 2001).

Limitations of the study

Finally, we would like to highlight some of the limitations of this study that we are aware of. First,

having more samples could give higher confidence to the results. Comparisons involving samples

from temperate climates would be equally interesting. Due to the quality of the gene annotations,

many genes could not be functionally interpreted. Moreover, due to the large number of differentially

expressed genes, we did not look in detail at all of the genes, and therefore, many interesting genes

may not have been highlighted in this study. While we tried to interpret our results based on model

species, primarily human studies, some of the interpretation may not be accurate. Particularly, most of

the studies are based on model species with more controlled settings, whereas the animals in our

study, especially Yakutian animals, live in harsh environmental conditions.

Conclusions

In conclusion, using blood as a starting material, we were able to capture more than 60% of cattle

genes to assess the differential gene expression in three breeds. The gene expression profiles in two

important indigenous cattle breeds that are adapted to the northern and even sub-arctic climate have

been reported for the first time. Gene expression comparisons between the three breed groups

revealed that Northern Finncattle are more closely related to Yakutian cattle than to Holstein cattle,

despite sharing similar living conditions with the Holstein cows that were analysed in this study.

Moreover, we identified breed-specific differences in maintaining immunity and disease resistance

mechanisms among the breeds included in our study. Some of our results, based on Holstein cattle,

explain how alternative biological processes can help imported animals cope with challenging

environmental conditions. Studies such as this one could identify genetic markers that may assist in

animal breeding and the sustainable utilization and conservation practices of animal genetic resources

in changing northern Eurasian environments.

Data availability

The raw sequence reads (FASTQ files) have been deposited to the European Nucleotide Archive

(ENA) under the accession number PRJEB28074 (please refer to Table 2 for sample specific

accessions). The script used during the differential expression analysis is available as supplementary

code (Code S1).

Acknowledgements:

We thank our colleague Tuula Marjatta Hamama from the Natural Resources Institute Finland (Luke)

for laboratory assistance and Innokenty Ammosov for valuable assistance in sampling in Yakutia.

This study belongs to the Arctic-Ark project, funded by the Academy of Finland (decision no.

286040). The authors wish to acknowledge the CSC – IT Center for Science, Finland, for

computational resources. This study was supported by the Finnish Functional Genomics Centre,

University of Turku and Åbo Akademi and Biocenter Finland. Special thanks to the owners of the

experimental animals for letting us collect the blood samples.

Author contributions:

J.K. conceived the analysis. J.K., M.H., T.R., J.P., H.L., R.P. and S.Z. participated in sample

collection. H.H. extracted mRNA from blood samples. K.P. and M.W.B. performed RNA-Seq data

analyses. K.P. wrote the manuscript. All authors read, revised and approved the final manuscript. The

authors declare that they have no competing interests.

References

Bai, X., Zheng, Z., Liu, B., Ji, X., Bai, Y., Zhang, W., 2016. Whole blood transcriptional profiling

comparison between different milk yield of Chinese Holstein cows using RNA-seq data. BMC

Genomics 17 Suppl 7, 512. https://doi.org/10.1186/s12864-016-2901-1

Berman, D.M., Wilkie, T.M., Gilman, A.G., 1996. GAIP and RGS4 are GTPase-activating proteins

for the Gi subfamily of G protein alpha subunits. Cell 86, 445–52.

Bindea, G., Mlecnik, B., Hackl, H., Charoentong, P., Tosolini, M., Kirilovsky, A., Fridman, W.-H.,

Pagès, F., Trajanoski, Z., Galon, J., 2009. ClueGO: a Cytoscape plug-in to decipher functionally

grouped gene ontology and pathway annotation networks. Bioinformatics 25, 1091–3.

https://doi.org/10.1093/bioinformatics/btp101

Blumer, K.J., 2004. The need for speed. Nature 427, 20–21. https://doi.org/10.1038/427020a

Cook, D.N., 1996. The role of MIP-1 alpha in inflammation and hematopoiesis. J. Leukoc. Biol. 59,

61–6.

Decker, J.E., Taylor, J.F., Kantanen, J., Millbrooke, A., Schnabel, R.D., Alexander, L.J., MacNeil,

M.D., 2016. Origins of cattle on Chirikof Island, Alaska, elucidated from genome-wide SNP

genotypes. Heredity (Edinb). 116, 502–505. https://doi.org/10.1038/hdy.2016.7

Dobin, A., Davis, C.A., Schlesinger, F., Drenkow, J., Zaleski, C., Jha, S., Batut, P., Chaisson, M.,

Gingeras, T.R., 2013. STAR: ultrafast universal RNA-seq aligner. Bioinformatics 29, 15–21.

https://doi.org/10.1093/bioinformatics/bts635

Durinck, S., Spellman, P.T., Birney, E., Huber, W., 2009. Mapping identifiers for the integration of

genomic datasets with the R/Bioconductor package biomaRt. Nat. Protoc. 4, 1184–91.

https://doi.org/10.1038/nprot.2009.97

Ebling, F.J.P., Barrett, P., 2008. The Regulation of Seasonal Changes in Food Intake and Body

Weight. J. Neuroendocrinol. 20, 827–833. https://doi.org/10.1111/j.1365-2826.2008.01721.x

Ewels, P., Magnusson, M., Lundin, S., Käller, M., 2016. MultiQC: summarize analysis results for

multiple tools and samples in a single report. Bioinformatics 32, 3047–3048.

https://doi.org/10.1093/bioinformatics/btw354

Gaouar, S.B.S., Da Silva, A., Ciani, E., Kdidi, S., Aouissat, M., Dhimi, L., Lafri, M., Maftah, A.,

Mehtar, N., 2015. Admixture and Local Breed Marginalization Threaten Algerian Sheep

Diversity. PLoS One 10, e0122667. https://doi.org/10.1371/journal.pone.0122667

Hediger, M.A., Romero, M.F., Peng, J.-B., Rolfs, A., Takanaga, H., Bruford, E.A., 2004. The ABCs

of solute carriers: physiological, pathological and therapeutic implications of human membrane

transport proteins. Pfl□gers Arch. Eur. J. Physiol. 447, 465–468.

https://doi.org/10.1007/s00424-003-1192-y

Hiyama, A., Skubutyte, R., Markova, D., Anderson, D.G., Yadla, S., Sakai, D., Mochida, J., Albert,

T.J., Shapiro, I.M., Risbud, M. V, 2011. Hypoxia activates the notch signaling pathway in cells

of the intervertebral disc: implications in degenerative disc disease. Arthritis Rheum. 63, 1355–

64. https://doi.org/10.1002/art.30246

Johnson, H., Scorrano, L., Korsmeyer, S.J., Ley, T.J., 2003. Cell death induced by granzyme C. Blood

101, 3093–3101. https://doi.org/10.1182/blood-2002-08-2485

Kaiserman, D., Bird, C.H., Sun, J., Matthews, A., Ung, K., Whisstock, J.C., Thompson, P.E., Trapani,

J.A., Bird, P.I., 2006. The major human and mouse granzymes are structurally and functionally

divergent. J. Cell Biol. 175, 619–30. https://doi.org/10.1083/jcb.200606073

Kantanen, J., Ammosov, I., Li, M.-H., Osva, A., Popov, R., 2009. A cow of the permafrost, in:

Granberg, L., Kantanen, J., Soini, K. (Eds.), Sakha Ynaga: Cattle of the Yakuts. Finnish
Academy of Science and Letters, pp. 19–44.

Kantanen, J., Løvendahl, P., Strandberg, E., Eythorsdottir, E., Li, M.-H., Kettunen-Prӓbel, A., Berg,
P., Meuwissen, T., 2015. Utilization of farm animal genetic resources in a changing agro-
ecological environment in the Nordic countries. Front. Genet. 6, 52.
https://doi.org/10.3389/fgene.2015.00052

Li, C., Cai, W., Zhou, C., Yin, H., Zhang, Z., Loor, J.J., Sun, D., Zhang, Q., Liu, J., Zhang, S., 2016.
RNA-Seq reveals 10 novel promising candidate genes affecting milk protein concentration in the
Chinese Holstein population. Sci. Rep. 6, 26813. https://doi.org/10.1038/srep26813

Li, M.-H., Kantanen, J., 2010. Genetic structure of Eurasian cattle (*Bos taurus*) based on
microsatellites: clarification for their breed classification. Anim. Genet. 41, 150–158.
https://doi.org/10.1111/j.1365-2052.2009.01980.x

Li, M.-H., Tapio, I., Vilkki, J., Ivanova, Z., Kiselyova, T., Marzanov, N., Ćinkulov, M., Stojanovic,
S., Ammosov, I., Popov, R., Kantanen, J., 2007. The genetic structure of cattle populations (Bos
taurus) in northern Eurasia and the neighbouring Near Eastern regions: implications for breeding
strategies and conservation. Mol. Ecol. 16, 3839–3853. https://doi.org/10.1111/j.1365-
294X.2007.03437.x

Love, M.I., Huber, W., Anders, S., 2014. Moderated estimation of fold change and dispersion for
RNA-seq data with DESeq2. Genome Biol. 15, 550. https://doi.org/10.1186/s13059-014-0550-8

MacDonald, G., Shi, L., Vande Velde, C., Lieberman, J., Greenberg, A.H., 1999. Mitochondria-
dependent and -independent regulation of Granzyme B-induced apoptosis. J. Exp. Med. 189,
131–44. https://doi.org/10.1084/JEM.189.1.131

Mirkena, T., Duguma, G., Haile, A., Tibbo, M., Okeyo, A.M., Wurzinger, M., Sölkner, J., 2010.
Genetics of adaptation in domestic farm animals: A review. Livest. Sci. 132, 1–12.
https://doi.org/10.1016/j.livsci.2010.05.003

Pokharel, K., Peippo, J., Honkatukia, M., Seppälä, A., Rautiainen, J., Ghanem, N., Hamama, T.-M.,

Crowe, M.A., Andersson, M., Li, M.-H., Kantanen, J., 2018. Integrated ovarian mRNA and

miRNA transcriptome profiling characterizes the genetic basis of prolificacy traits in sheep

(Ovis aries). BMC Genomics 19. https://doi.org/10.1186/s12864-017-4400-4

Politi, K., Feirt, N., Kitajewski, J., 2004. Notch in mammary gland development and breast cancer.

Semin. Cancer Biol. 14, 341–347. https://doi.org/10.1016/j.semcancer.2004.04.013

Rocha, P.L., Branco, L.G., 1998. Seasonal changes in the cardiovascular, respiratory and metabolic

responses to temperature and hypoxia in the bullfrog Rana catesbeiana. J. Exp. Biol. 201, 761–8.

Russell, J.H., Ley, T.J., 2002. Lympphocyte-mediated cytotoxicity. Annu. Rev. Immunol. 20, 323–

370. https://doi.org/10.1146/annurev.immunol.20.100201.131730

Sandri, M., Stefanon, B., Loor, J.J., 2015. Transcriptome profiles of whole blood in Italian Holstein

and Italian Simmental lactating cows diverging for genetic merit for milk protein. J. Dairy Sci.

98, 6119–6127. https://doi.org/10.3168/JDS.2014-9049

Shannon, P., Markiel, A., Ozier, O., Baliga, N.S., Wang, J.T., Ramage, D., Amin, N., Schwikowski,

B., Ideker, T., 2003. Cytoscape: a software environment for integrated models of biomolecular

interaction networks. Genome Res. 13, 2498–504. https://doi.org/10.1101/gr.1239303

Silva-Vignato, B., Coutinho, L.L., Cesar, A.S.M., Poleti, M.D., Regitano, L.C.A., Balieiro, J.C.C.,

2017. Comparative muscle transcriptome associated with carcass traits of Nellore cattle. BMC

Genomics 18, 506. https://doi.org/10.1186/s12864-017-3897-x

Simon Andrews, n.d. FastQC A Quality Control tool for High Throughput Sequence Data [WWW

Document]. http://www.bioinformatics.babraham.ac.uk/projects/fastqc/. URL

http://www.bioinformatics.babraham.ac.uk/projects/fastqc/ (accessed 1.18.13).

Smedley, D., Haider, S., Ballester, B., Holland, R., London, D., Thorisson, G., Kasprzyk, A., 2009.

BioMart--biological queries made easy. BMC Genomics 10, 22. https://doi.org/10.1186/1471-

2164-10-22

Soreq, H., Seidman, S., 2001. Acetylcholinesterase — new roles for an old actor. Nat. Rev. Neurosci.

2, 294–302. https://doi.org/10.1038/35067589

Wang, Z., Gerstein, M., Snyder, M., 2009. RNA-Seq: a revolutionary tool for transcriptomics. Nat.

Rev. Genet. 10, 57–63. https://doi.org/10.1038/nrg2484

Wu, N.C., Cramp, R.L., Franklin, C.E., 2017. Living with a leaky skin: upregulation of ion transport

proteins during sloughing. J. Exp. Biol. 220, 2026–2035. https://doi.org/10.1242/jeb.151738

Yalcin-Ozuysal, Ö., Fiche, M., Guitierrez, M., Wagner, K.-U., Raffoul, W., Brisken, C., 2010.

Antagonistic roles of Notch and p63 in controlling mammary epithelial cell fates. Cell Death

Differ. 17, 1600–1612. https://doi.org/10.1038/cdd.2010.37

Yang, J., Vrettou, C., Connelley, T., Morrison, W.I., 2018. Identification and annotation of bovine

granzyme genes reveals a novel granzyme encoded within the trypsin-like locus.

Immunogenetics 1–13. https://doi.org/10.1007/s00251-018-1062-6

Yurchenko, A., Yudin, N., Aitnazarov, R., Plyusnina, A., Brukhin, V., Soloshenko, V., Lhasaranov,

B., Popov, R., Paronyan, I.A., Plemyashov, K. V., Larkin, D.M., 2018. Genome-wide

genotyping uncovers genetic profiles and history of the Russian cattle breeds. Heredity (Edinb).

120, 125–137. https://doi.org/10.1038/s41437-017-0024-3

Zhang, P., Mende, U., 2014. Functional role, mechanisms of regulation, and therapeutic potential of

regulator of G protein signaling 2 in the heart. Trends Cardiovasc. Med. 24, 85–93.

https://doi.org/10.1016/j.tcm.2013.07.002

Zimin, A. V, Delcher, A.L., Florea, L., Kelley, D.R., Schatz, M.C., Puiu, D., Hanrahan, F., Pertea, G.,

Van Tassell, C.P., Sonstegard, T.S., Marçais, G., Roberts, M., Subramanian, P., Yorke, J.A.,

Salzberg, S.L., 2009. A whole-genome assembly of the domestic cow, Bos taurus. Genome Biol.

10, R42. https://doi.org/10.1186/gb-2009-10-4-r42

Zinovieva, N.A., Dotsev, A.V., Sermyagin, A.A., Wimmers, K., Reyer, H., Sölkner, J., Deniskova,

T.E., Brem, G., 2016. Study of genetic diversity and population structure of five Russian cattle

breeds using whole-genome SNP analysis. Sel'skokhozyaistvennaya Biol. 51, 788–800.

https://doi.org/10.15389/agrobiology.2016.6.788eng

Figures

453
454

455   Fig. 1: Co-expression of the top 500 genes among the three breed groups.

positive regulation of natural killer cell chemotaxis 49.17% **

regulation of circadian rhythm 0.83% **
protein self-association 0.83% **
granzyme-mediated apoptotic signaling pathway 2.5% **
cellular response to tumor necrosis factor 3.33% **

negative regulation of cytokine production 7.5% **

negative regulation of viral process 18.33% **

regulation of phagocytosis 17.5% **

a

interaction with symbiont 5.26% **
substrate adhesion-dependent cell spreading 5.26% **

gas transport 5.26% **
peptide cross-linking 5.26% **
osteoclast differentiation 5.26% **

regulation of endothelial cell proliferation 10.53% **

regulation of lipid transport 21.05% **

regulation of protein maturation 10.53% **

erythrocyte differentiation 15.79% **

regulation of cytokine-mediated signaling pathway 15.79% **

b

Fig. 2: A functional annotation of the significantly differentially expressed genes in Yakutian cattle and Northern Finncattle. The GO terms associated with upregulated genes in Yakutian cattle (a) and Northern Finncattle (b). Similar GO terms were grouped together, and thus, the percentages at the ends of each representative GO term indicate the total percentages of linked GO terms. The significant GO terms at the p<0.001 statistical level are indicated by double (**) asterisks.

chemokine-mediated signaling pathway 28.97% **

cell-cell junction assembly 0.93% .
regulation of circadian rhythm 0.93%
protein self-association 0.93%
dendritic cell differentiation 0.93% *
regulation of glucose metabolic process 0.93% *
megakaryocyte differentiation 0.93% *
multicellular organismal response to stress 0.93%
homotypic cell-cell adhesion 0.93%
positive regulation of angiogenesis 1.87% .
G-protein coupled purinergic receptor signaling pathway 1.87% .
regulation of phagocytosis 1.87% .
regulation of muscle cell differentiation 1.87%
cation-transporting ATPase activity 2.8%
granzyme-mediated apoptotic signaling pathway 3.74% **
protein kinase activator activity 4.67% *
production of molecular mediator involved in inflammatory response 5.61% *
regulation of T cell activation 9.35% *

regulation of natural killer cell mediated immunity 15.89% **

kinase activator activity 14.02% *

a

Cytokine-cytokine receptor interaction 8.33% **

Cell adhesion molecules (CAMs) 8.33% *

Cytosolic DNA-sensing pathway 8.33% *

Natural killer cell mediated cytotoxicity 58.33% **

Endocrine and other factor-regulated calcium reabsorption 8.33% *

African trypanosomiasis 8.33% *

b

inflammatory response 23.21% **

negative regulation of leukocyte activation 1.79% *
positive regulation of canonical Wnt signaling pathway 1.79% *
negative regulation of protein maturation 1.79% **
wound healing 3.57% **
positive regulation of endocytosis 3.57% *
negative regulation of cytokine production 5.36% **
regulation of myeloid cell differentiation 10.71% *

regulation of neutrophil chemotaxis 21.43% **

secondary active transmembrane transporter activity 12.5% **

regulation of response to external stimulus 14.29% **

c

Non-small cell lung cancer 20.69%

ABC transporters 3.45%
Complement and coagulation cascades 3.45%
Pancreatic secretion 3.45%
Protein digestion and absorption 3.45%
Chemokine signaling pathway 3.45%
Mineral absorption 3.45%
Cocaine addiction 3.45%
Prostate cancer 3.45%
p53 signaling pathway 3.45%
Osteoclast differentiation 3.45%
NOD-like receptor signaling pathway 3.45% *
Hematopoietic cell lineage 3.45%
Adipocytokine signaling pathway 3.45%
Bladder cancer 6.9%

Legionellosis 13.79%

Notch signaling pathway 6.9% .

Cytokine-cytokine receptor interaction 6.9% **

d

463
464

Fig. 3: A functional annotation of the genes that were significantly differentially expressed between

Yakutian cattle and Holstein cattle. (a) The GO terms and (b) KEGG pathways that were associated

with upregulated genes in Yakutian cattle. (c) The GO terms and (d) KEGG pathways that were

associated with upregulated genes in Holstein cattle. Similar GO terms and KEGG pathways were

grouped together, and thus, the percentages at the ends of each term indicate the total percentages of

linked GO terms and KEGG pathways. The significant GO terms and KEGG pathways at the p<0.05

and p<0.001 statistical levels are indicated by single (*) and double (**) asterisks, respectively.

472
473

474   Fig. 4: A functional annotation of the genes that were upregulated in Holstein cattle compared to

475   Northern Finncattle. Similar GO terms were grouped together, and thus, the percentages at the ends of

476   each term indicate the total percentages of linked GO terms. The significant GO terms at the p<0.001

477   statistical level are indicated by double (**) asterisks.

478
479

Fig 5: A functional annotation of the significantly differentially expressed genes between Yakutian

and Holstein cattle. Nodes represent GO terms, with node size corresponding to the significance of

term enrichment, and functionally related groups partially overlap. A network of (a) GO terms and (b)

KEGG pathways that are associated with the genes upregulated in Yakutian cattle. A network of (c)

GO terms and (d) KEGG pathways that are associated with upregulated genes in Holstein cattle. The

representative GO terms and KEGG pathways are highlighted in bold.

Tables

Table 1: Sample summary.

Sample ID	Breed	ENA Accession	No. of reads (million)	Uniquely mapped reads %	No. of genes expressed
FC7	Northern Finncattle	ERS2639473	49.4	87.2	12,106
FC8	Northern Finncattle	ERS2639474	107.6	87.1	15,038
FC9	Northern Finncattle	ERS2639475	89.4	89.7	14,811
HC1	Holstein	ERS2639476	104.4	91.7	14,768

HC2	Holstein	ERS2639477	62.8	91.0	14,140
HC3	Holstein	ERS2639478	62.4	92.4	13,831
YC6	Yakutian	ERS2639479	69.2	87.3	14,047
YC10	Yakutian	ERS2639480	69.8	88.2	13,736
YC11	Yakutian	ERS2639481	65.8	88.5	14,371

489 Table 2: Numbers of uniquely differentially expressed genes in different breeds.

Breed	No. of uniquely differentially expressed genes	Upregulated genes	Downregulated genes
Yakutian cattle	139	89	52
Northern Finncattle	9	2	7
Holstein cattle	200	162	48

Supplementary files

Fig. S1: Venn diagram based on the list of significantly differentially expressed genes following all

three possible comparisons

Fig. S2: Figure showing the closely located Granzyme B in the genome

Table S1: Table listing the expression levels of each sample

Table S2: List of significantly differentially expressed genes between Yakutian cattle and Northern

Finncattle

Table S3: The GO terms associated with upregulated genes in Yakutian cattle when compared with

Northern Finncattle

Table S4: The KEGG pathways associated with upregulated genes in Yakutian cattle when compared

with Northern Finncattle

Table S5: The GO terms associated with upregulated genes in Northern Finncattle when compared

with Yakutian cattle

Table S6: Summary of significantly differentially expressed genes between Yakutian cattle and

Holstein cattle