An equine Endothelin 3 cis-regulatory variant links blood pressure

modulation to elite racing performance

Endothelin 3 affects blood pressure regulation and performance in horses

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Abstract

A previous selective sweep analysis of horse racing performance revealed a 19.6 kb candidate

region approximately 50 kb downstream of the Endothelin 3 (*EDN3*) gene. EDN3 and other

endothelin family members are associated with blood pressure regulation in humans and other

species, but similar association studies in horses are lacking. We hypothesized that the sweep

region includes a regulatory element acting on *EDN3* transcription, ultimately affecting blood

pressure regulation and athletic performance in horses. Selective sweep fine-mapping

identified a 5.5 kb haplotype of 14 SNPs shared within Coldblooded trotters (CBT) and

Standardbreds (SB). Most SNPs overlapped potential transcription factor binding sites, and

haplotype analysis showed significant association with all tested performance traits in CBTs

and earnings in SBs. From those, two haplotypes were defined: an elite performing haplotype

(EPH) and a sub-elite performing haplotype (SPH). While the majority of SNPs in the

haplotype were part of the standing variation already found in pre-domestication horses, there

has been an increase in the frequencies of the alternative alleles during the whole history of

horse domestication. Horses homozygous for EPH had significantly higher plasma levels of

EDN3, lower levels of EDN1, and lower exercise-related blood pressure compared to SPH

homozygous horses. Additionally, a global proteomic analysis of plasma from EPH or SPH

homozygous horses revealed higher levels of proteins involved in pathways related to immune

response and complement activation in the SPH horses. This is the first study to demonstrate

an association between the *EDN3* gene, blood pressure regulation, and athletic performance in

horses. The results advance our understanding of the molecular genetics of athletic performance, exercise-related blood pressure regulation, and biological processes activated by intense exercise.

Keywords: Athletic performance, Blood pressure, Endothelin, *Equus caballus*, Exercise, Genetics, Harness racing, Horse, Trotting

Author summary

The horse is one of the most common species used for studying athletic performance. For centuries, horses have been used by humans for transportation, agriculture and entertainment and this has resulted in selection for various traits related to athletic performance. A previous study discovered that a genetic region close to the *Endothelin3* gene was associated with harness racing performance. Endothelin3 is known to be involved in blood pressure regulation and therefore we hypothesized that this region influences blood pressure and racing performance in horses. In this study we have used additional horses and fine-mapped the candidate region and we also measured blood pressure in Coldblooded trotters during exercise. Horses with two copies of the elite-performing haplotype had higher levels of Endothelin3 in plasma, lower blood pressure and better racing performance results, compared to horses with two copies of the sub-elite performing haplotype. We also discovered that horses with the sub-elite performing haplotype had higher levels of proteins related to the immune system in plasma. This study is the first to link Endothelin3 to blood pressure regulation and performance in horses. It broadens the understanding of the biological mechanisms behind blood pressure regulation as well as inflammation and coagulation system in relation to racing performance.

Background

The use of domestic animals as models for genomic research has provided not only basic knowledge concerning gene function and biological mechanisms but also a complementary view on genotype–phenotype relationships (1). The horse is one of the most popular species used to study athletic performance. They have been intensively selected for centuries, to engender the optimal physical capacity for strength, speed, and endurance (2,3). Additionally, their recent population history, involving closed populations with similar phenotypic traits within breeds and large variations across breeds, has created a favorable genome structure for genetic mapping. These factors combined make the horse an optimal model for studying the molecular genetics underlying athletic performance and the complex biological processes activated by exercise (1,2,4,5). Previous work in horses has started to reveal the effect of genes linked to complex traits such as muscle mass and locomotion patterns on athletic performance, and many potential candidate genetic variants have been identified (6–12). However, understanding the mechanisms by which the identified variants exert their functions is much more challenging.

A selective sweep study on harness racing breeds revealed a 19.6 kb genomic region on chromosome 22, located about 50 kb downstream of the *Endothelin 3* (*EDN3*) gene, under selection in Coldblooded trotters (CBTs) (13). Five SNPs in high linkage disequilibrium (LD $r^2 = 0.92$–0.94) were significantly associated with racing performance, including the number of victories, earnings, and racing times. SNP NC_009165.3:g.46717860C>T (EquCab3.0) showed the strongest association within the sweep and was further genotyped in 18 additional horse breeds. The favorable T allele was found in high frequency in breeds used for racing, while it generally remained at low frequency in ponies and draught horses. It was hypothesized

that the identified region might contain a regulatory element influencing either the expression

of *EDN3* or other genes nearby (13).

The endothelin system encompasses the EDN ligands (e.g., EDN-1 to -3) and their G protein-

coupled receptors (endothelin receptor type A and B). Together, they are involved in a broad

range of biological functions, including cell proliferation and vasoconstriction (14). EDN2 and

-3 can be considered as natural antagonists for EDN1 (15). The endothelin receptor type A

(EDNRA) has the highest affinity for EDN1 and EDN2 but a low affinity for EDN3 (16). The

endothelin receptor type B (EDNRB) has equal affinity for all three isoforms of endothelin

(17). During exercise, endothelins are involved in the dynamic redistribution of blood flow to

the lung and active muscles, limiting the flow to non-exercise-essential tissues, such as kidneys

and intestines (15). Generally, EDNRA activation causes vasoconstriction and increased blood

pressure, while EDNRB activation results in nitric oxide release, vasodilatation, and decreased

blood pressure (18).

To date, most studies of the EDN system have focused on the role of EDN1. In human health

studies, *EDN1* genetic variants were found to be implicated in hypertension (19). In

performance trials, increased EDN1 plasma levels were observed in relation to exercise

duration (20), with elevated blood pressure leading to impaired exercise capacity in elite

athletes (20,21). There have also been performance studies in horses, where EDN1 plasma

levels largely mirrored the results found in humans, showing that the EDN1 response to

exercise challenge is conserved across both species (22). There is a clear role for

vasoconstriction in exercise performance, but the action of the EDN3 ligand is unclear. In this

study, we aimed to fill this gap of knowledge. First through the fine-mapping of the *EDN3*

region flanking the selective sweep, then the characterization of the spatial and temporal

distribution of key genetic variants within horse populations, and finally by evaluating the

effect of opposite haplotypes on blood pressure, as well as EDN1 and EDN3 plasma levels,

before and during exercise. We also performed a global proteomic study on plasma to identify

differentially expressed proteins amongst the various *EDN3* haplotypes. While the key breed

of the study was the CBT, a popular harness racing breed in Sweden and Norway, the results

are placed in the context of additional racing breeds. The results presented here on the

endothelin pathway not only give new insights into genetic regulation of its established

vasomodulatory properties, but also its emerging role in inflammation.

Results

Fine-mapping and comparative analyses suggest a regulatory role of the selected

region

A 2018 study using the sparse Illumina SNP50 Genotyping BeadChip, identified a 19.6 kb

selective sweep shared between SBs and CBTs (ECA22:46,702,297-46,721,892; Figure 1A).

The sweep was located between *EDN3* and ENSECAG00000039543 (lncRNA), and five SNPs

within this region were found to be associated with harness racing performance (13). From

those five SNPs, g.46717860 C>T was the most significant. With a four-step process, we

reduced this 19.6 kb sweep region to a minimal 5.5 kb region (Figure 1).

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In step one, we performed a sweep-performance association analysis using additional horse

material (n=661) and SNP genotypes extracted from the 670K Axiom Equine Genotyping

Array (Figure 1A). However, supplementing the 400 CBTs used originally in the 2018

study(13) with 221 additional CBTs did not provide increased resolution into the 19.6 kb

sweep. Seven SNPs within the original sweep span were extracted from the array, including

g.46717860. However, pairwise LD ($r^2 > 0.6$) to g.46717860 resulted in the same five SNP set

as found in (13), with all SNPs remaining significantly associated with harness racing

performance traits in linear models (i.e., number of victories, earnings, and race times, $P \leq$

0.05) (Supplementary Table S1).

Figure 1. A. Overview of the 19.6 kb 2018 selective sweep region including all annotated genes in the

region (13). B. Zoomed in view of the 2018 selective sweep region including the 17 SNPs used in the

haplotype analysis. C. Overview of the 5.5 kb minimum shared haplotype within the Coldblooded

trotters and Standardbreds, including all SNPs from the haplotype analysis. D. Horse functional data as

generated by chip-seq of histone modifications from hoof lamina (from Equine Functional Annotation

of Animal Genomes, FAANG) indicating an active enhancer in our region of interest. In addition, ~ 2

kb of the associated haplotype region aligned with the human reference, including segments of an

ORegAnno regulatory element, pinpointing a curated regulatory annotation of this region,

(OREG1514229), GTEx cis-eQTL variants regulating *EDN3* in esophagus mucosa tissue.

In step two, we performed variant discovery to increase variant density across the 19.6 kb

sweep region. Here, Illumina short-read whole genome sequences (WGSs) were generated

from two CBTs and two SBs (the most common breed used for harness racing). In each breed,

the horses were selected to be homozygous for different alleles at SNP g.46717860 (CC and

TT, respectively) and to have either high (TT) or low (CC) earnings per start (Supplementary

Table S2). From the 19.6 kb sweep region, 78 SNPs and six indels (one 400 bp deletion, one

23 bp deletion, three single bp deletions, and one single bp insertion) were identified. To

possibly detect new structural variants, targeted long-read sequencing using Oxford Nanopore

technology (ONT) was performed across the sweep region for eight horses: four CBTs with

g.46717860-TT genotype and high performance, and four CBTs with the opposite features,

including the two CBTs used for WGS (Supplementary Table S2). However, no new structural

genetic variants were identified.

In step three, we prioritized variants for further analysis. While the whole genome sequenced

elite CBT (high earnings per start) was homozygous for the 400 bp deletion

(ECA22:46,714,602-46,715,003), the elite SB was heterozygous, and the deletion was not

detected in the low-performing (low earnings per start) horses. The deletion was, thus, tested

for association with racing performance traits in 479 CBTs using linear models, but the results

were not significant (Supplementary Table S3). From the remaining variants identified, a total

of 24 SNPs and one single base pair deletion, evenly distributed within the region, were

selected (see methods) and genotyped in four donkeys and 412 horses sampled from 13

different breeds. SNP g.46717860 was not included in this set but was genotyped separately in

170 horses and imputed in the remainder. At this stage, 24 SNPs and 407 horses passed quality

control measures. Pairwise LD estimation with SNP g.46717860 revealed 15 SNPs in strong

LD ($r^2 \geq 0.53$, 14 SNPs with an r^2 score > 0.98) (ECA22: 46,708,983-46,719,042). LD decayed

outside the region ($r^2 < 0.07$), leaving 17 SNPs (15 SNPs in LD and two flanking SNPs) for

haplotype analysis (Figure 1B). Phasing revealed 15 different haplotypes, five showing a

frequency > 2 %. All haplotypes with a frequency > 5 % in the total sample set or within each

breed are presented in Table 1 (disregarding breeds represented by less than five horses).

Table 1. Haplotype frequencies in different breeds

Haplotype (ECA22: 46,708,110-46,719,906)	All breeds (n=411)	CBT (n=222)	Arabians (n=29)	TB (n=30)	SB (n=45)	Ardennes (n=20)	NSH (n=18)	Exmoor (n=17)	Icelandic (n=7)
1) GCAAGCGTTTCTCAAAA	0.30	0.27	0.74	0.68	0.36	0.00	0.03	0.00	0.00
2) GTGGTTACCCTCTGGGA	0.28	0.25	0.00	0.07	0.00	0.60	0.86	1.00	0.00
3) ACAAGCGTTTCTCAAAG	0.27	0.32	0.26	0.25	0.41	0.10	<0.01	0.00	0.00
4) ACGGTTACCCTCTGGGG	0.11	0.14	0.00	0.00	0.20	0.00	0.01	0.00	0.43
5) ACGGTTACCCTCTGGGA	0.03	0.00	0.00	0.00	0.04	0.18	0.03	0.00	0.36
6) ACAAGCGTTTCTCGAAA	<0.01	0.00	0.00	0.00	0.00	0.10	0.00	0.00	0.00
7) ACGGGCGTTCCTCAAAG	<0.01	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.07
8) GCAAGCGTTTCTCAAGG	<0.01	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.07
9) GCAAGTGTTTCTCAAGG	<0.01	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.07

[1] Order of SNPs. Chr 22: 46708110, 46708983, 46713478, 46715974, 46717451, 46717528,

46717742, 46717854, 46717860, 46717999, 46718095, 46718361, 46718436, 46718895, 46718964,

46719042, 46719906

CBT: Coldblooded trotter, TB: Thoroughbred, SB: Standardbred, NSH: North-Swedish draught horse

The association between haplotype and harness racing performance traits was performed in

CBTs (n=180) and SBs (n=38) using generalized linear model regression (GLM) analysis in

the statistical software R (23) (Table 2 and 3). Only haplotypes with a frequency > 2 % in the

population were included in the analysis. Of the four haplotypes in CBTs, haplotype 3 (H3),

carrying the g.46717860-T high performance associated allele, was the most common (0.32)

and was defined as the base haplotype. Two haplotypes, H2 and H4, respectively, demonstrated

a significant negative effect on all performance traits tested, and each carried the g.46717860-

C low performance associated allele (Table 2). Four haplotypes were also available for testing

within SBs with performance data. Three haplotypes were shared with CBTs (H1, H3 and H4)

and one (H5) was shared with other breeds (Table 3). As for the CBTs, H3 was the most

common haplotype (0.43) and was defined as the base haplotype in this analysis. Two

haplotypes demonstrated significant negative effects on performance traits: H4 (earnings and

earnings per start) and H5 (all traits tested). The above analysis revealed a minimum 5,564 bp,

14 SNP shared haplotype, significantly associated with racing performance within CBTs

(ECA22: 46,713,478-46,719,042). The same variants were also significantly associated with

racing performance in SBs. This allowed for the definition of an elite-performing haplotype

(EPH: AAGCGTTTCTCAAA), and a sub-elite-performing haplotype (SPH:

GGTTACCCTCTGGG) (Supplementary Table S4) (Figure 1C). Of note, the reference

EquCab3.0 genome represents the SPH variant.

222 Table 2. Haplotype frequencies, haplotype coefficient and *P*-values for the performance analysis in

223 Coldblooded trotters (n=180)

	Haplotype[2] (H3)		Haplotype (H1)		Haplotype (H2)		Haplotype (H4)	
Haplotype[1]	ACAAGCGTTTCTCAAAG		GCAAGCGTTTCTCAAAA		GTGGTTACCCTCTGGGA		ACGGTTACCCTCTGGGG	
Frequency	0.32		0.27		0.27		0.14	
	Haplotype coefficient	P-value[3]	Haplotype coefficient	P-value[3]	Haplotype coefficient	P-value[3]	Haplotype coefficient	P-value[3]
No. of starts[5]	-	-	-0.30	0.43	-1.00	**0.004**	-1.05	**0.02**
No. of wins[5]	-	-	-0.04	0.31	-0.14	**<0.001**	-0.17	**<0.001**
No. of placings[5]	-	-	-0.06	0.13	-0.12	**<0.001**	-0.14	**0.002**
Earnings (SEK)[4,5]	-	-	-0.31	0.05	-0.53	**<0.001**	-0.56	**0.002**
Earnings/start[5]	-	-	-0.19	0.07	-0.45	**<0.001**	-0.42	**<0.001**
Best time (sec/km)[5]	-	-	0.03	0.16	0.07	**<0.001**	0.08	**0.002**

[1]Order of SNPs. Chr 22: 46708110, 46708983, 46713478, 46715974, 46717451, 46717528,

46717742, 46717854, 46717860, 46717999, 46718095, 46718361, 46718436, 46718895, 46718964,

46719042, 46719906

[2]The most common haplotype was used as the base haplotype

[3]A GLM regression analysis was performed in R. Sex, age, birth country, number of starts (when

applicable) and *DMRT3* genotype were included as fixed effects. Significant results (P≤ 0.05) in bold

[4] SEK, Swedish kronor

[5] Log transformed values (log10 +1) were used for wins and placings. Number of starts were square

root transformed. Earnings and earnings per start were transformed applying ln(Earnings + 1000).

CBT: Coldblooded trotter, TB: Thoroughbred, SB: Standardbred, NSH: North-Swedish draught horse

	Haplotype[2] (H3)		Haplotype (H1)		Haplotype (H4)		Haplotype (H5)	
Haplotype[1]	ACAAGCGTTTCTCAAAG		GCAAGCGTTTCTCAAAA		ACGGTTACCCTCTGGGG		ACGGTTACCCTCTGGGA	
Frequency	0.43		0.34		0.18		0.04	
	Haplotype coefficient	P-value[3]	Haplotype coefficient	P-value[3]	Haplotype coefficient	P-value[3]	Haplotype coefficient	P-value[3]
No. of starts[5]	-	-	0.12	0.82	0.48	0.41	-0.61	**0.003**
No. of wins[5]	-	-	-0.02	0.74	-0.08	0.35	-0.47	**<0.001**
No. of placings[5]	-	-	-0.03	0.63	-0.11	0.06	-0.25	**<0.001**
Earnings (SEK)[4, 5]	-	-	-0.17	0.53	-1.42	**<0.001**	-1.32	**<0.001**
Earnings/start[5]	-	-	-0.10	0.66	-1.14	**<0.001**	-1.12	**<0.001**
Best time (sec/km)[5]	-	-	-0.10	0.37	0.15	0.24	0.65	**0.01**

[1]Order of SNPs. Chr 22: 46708110, 46708983, 46713478, 46715974, 46717451, 46717528,

46717742, 46717854, 46717860, 46717999, 46718095, 46718361, 46718436, 46718895, 46718964,

46719042, 46719906

[2]The most common haplotype was used as the base haplotype

[3]A GLM regression analysis was performed in R. Sex, age, birth country and number of starts (when

applicable) were included as fixed effects. Significant results (P≤ 0.05) in bold

[4] SEK, Swedish kronor

[5] Log transformed values (log10 +1) were used for wins and placings. Number of starts were square

root transformed. Earnings and earnings per start were transformed applying ln(Earnings + 1000).

CBT: Coldblooded trotter, TB: Thoroughbred, SB: Standardbred, NSH: North-Swedish draught horse

In step four, we assessed the potential functional impact of the minimal 5.5 kb region and each

of the 14 SNP haplotype variants. Using horse functional data from two Thoroughbreds (from

the Equine Functional Annotation of Animal Genomes, FAANG), we found three SNPs

(g.46717860, g.46718095 and g.46718113) that overlapped potential active enhancers, i.e.,

H3K4me1 and H3K27ac marks (24,25) (Figure 1D). Both Thoroughbreds were homozygous

for EPH, with the histone marks indicating the presence of an active enhancer in the hoof

lamina, present but less active in lung and ovarian tissue, and repressed in the skin(24,25). In

comparative analyses, a genome liftover to hg38 indicated that, although not well conserved,

~ 2 kb of the associated haplotype region aligned with the human reference, including segments

of an ORegAnno regulatory element (OREG1514229), GTEx cis-eQTL variants regulating

EDN3 in esophagus mucosa tissue (Figure 1D). These combined regulatory signals prompted

us to test the potential for transcription factor binding affinity changes in relation to the EPH

and SPH alleles. Most SNP allele changes (12 out of 14 SNPs) altered transcription factor

binding (Supplementary Table S4). For EPH, this meant the perturbation of matrices for stress

responders, e.g., IRF2 and NFKB1, while creating sites for developmental transcription factors

such as NR4A2 and HOXA5.

The elite performing haplotype has an additive effect on performance traits.

Given that all 14 associated variants within the minimum shared 5.5 kb haplotype were in

perfect LD, we used the genotypes at g.46718095 as a proxy for EPH (C allele) and SPH (T

allele) haplotypic pairs and re-assessed the association between haplotypes and racing

performance. For CBTs (n=539), EPH homo- or heterozygotes outperformed the SPH

homozygotes (Table 4). In SBs, there were no statistically significant differences between the

three genotypes (Table 5). However, when analyzing horses carrying at least one T allele versus

CC horses, the results were significant for several performance traits (Table 5), including the

number of wins, placings and earnings. However, the number of wins and placings was higher

in horses with at least one T allele, while the CC horses earned more money than the TT/TC

horses (Table 5).

Table 4. Performance results in 539 Coldblooded trotters for SNP g.46718095 T>C

Genotype	TT (n=65)		TC (n=256)		CC (n=218)		P[2]		
Performance trait[1]	Mean	Median	Mean	Median	Mean	Median	TT/TC	TT/CC	TC/CC
No. of starts	31.9	22.0	40.2	28.0	41.3	31.0	**0.01**	**0.02**	1.00
No. of wins	2.8	1.0	5.7	3.0	5.9	3.0	**0.003**	**0.02**	0.65
No. of placings (1-3)	9.3	4.0	14.2	9.0	14.5	9.0	**0.005**	0.07	0.38
Earnings (SEK)	207,295	51,772	438,700	140,161	453,911	115,312	**0.006**	**0.05**	0.58
Earnings per start (SEK)	4,752	2,976	9,666	4,362	8,469	3,771	**0.003**	0.08	0.20
Best km time (sec)[3]	90.9	91.4	88.5	88.3	88.9	88.9	**0.001**	**0.02**	0.44

[1]Log transformed values (log10 +1) were used for wins and placings. Earnings and earnings per start

were transformed applying ln(earnings + 1000). Square root transformations were used for number of

starts

[2]A linear model was performed in R. Sex, age, birth country, number of starts and *DMRT3* genotype

were included as fixed effects (when significant). Least square means and Tukey were used as post

hoc tests. Significant results ($P \leq 0.05$) in bold

[3]Only races where the horse did not gallop were included

301   Table 5. Performance results in 271 Standardbreds for SNP g.46718095 T>C

Genotype	TT (n=7)		TC (n=62)		CC (n=202)		P[2]			
Perf. trait[1]	Mean	Median	Mean	Median	Mean	Median	TT/TC	TT/CC	TC/CC	CC/ TT+TC
No. of starts	43.0	48.0	42.9	38.0	35.6	31.0	0.85	0.60	0.52	0.20
No. of wins	6.7	5.0	6.3	4.0	5.7	5.0	0.82	0.99	0.10	**0.04**
No. of placings (1-3)	15.6	12.0	16.1	16.0	13.7	11.0	0.86	0.96	0.08	**0.03**
Earnings (SEK)	700,008	322,300	730,124	332,300	1,075,529	369,675	0.92	0.43	0.06	**0.01**
Earnings per start (SEK)	11,981	5,463	16,673	9,765	35,346	9,897	0.86	0.44	0.16	**0.04**
Best km time (sec)[3]	73.1	73.9	73.3	72.8	73.1	72.9	0.85	1.00	0.19	0.10

[1]Log transformed values (log10 +1) were used for wins and placings. Earnings and earnings per start

were transformed applying ln(earnings + 1000). Square root transformations were used for number of

starts

[2]A linear model was performed in R. Sex, age, birth country and number of starts were included as

fixed effects (when significant). Least square means and Tukey were used as post hoc tests.

Significant results ($P \leq 0.05$) in bold

[3]Only races where the horse did not gallop were included

Our racing performance analyses revealed that the g.46718095 genotypes did not distribute

according to HWE. This prompted us to ask if there had been selection on this position or

perhaps the minimal haplotype in other breeds. While each individual's performance status is

unknown, a trend for increased frequency of the C allele in traditional performance breeds was

observed, e.g., Arabian horses, Thoroughbreds and Warmbloods. In contrast, this allele was

low or absent in draft horses and ponies (Table 6).

Table 6. Allele frequency of SNP g.46718095 T>C

Breed	n	T	C
Arabian horses	30	0.00	1.00
Thoroughbred	30	0.07	0.93
Standardbred	306	0.14	0.86
Warmblood	3	0.17	0.83
Coldblooded trotters	539	0.36	0.64
Finnhorses	4	0.50	0.50
Gotland pony	2	0.50	0.50
Shetland pony	18	0.64	0.36
Ardennes	20	0.78	0.22
Icelandic horses	11	0.86	0.14
North-Swedish draught	25	0.88	0.12
Exmoor pony	23	1.00	0.00
Fjord horse	1	1.00	0.00
Przewalski	4	1.00	0.00

Increased frequency of the favorable haplotype coincides with the time of horse domestication

For all 14 SNPs within the minimum shared haplotype in the trotters, we investigated the allele frequency over time using the mapDATAge package (26). For all SNPs but two (SNP 6 and 7 in Figure 2) there was an increase of the alternate allele from 7,500-5,500 years ago (Figure 2). For all SNPs considered, the two alleles segregated in specimens pre-dating the rise and spread of the DOM2 genetic lineage of modern domestic horses (data not shown). The dataset considered included 431 ancient genomes previously characterized (2,27–31).

Figure 2. Frequency of the alternative allele over time from the mapDATAge package (26) for all 14

SNPs within the minimum shared haplotype in the trotters. The orange dots represent the number of

horse samples considered in a given time bin of 2,000 years, and the blue line represents the

frequency for the alternative allele estimated from allele counts. Figure created with Biorender.com.

Elite performance haplotype linked to lower blood pressure

To link genotype to phenotype, we assessed the physiological difference between CBTs

homozygous for SPH (n=5) and EPH (n=13) before, during and post-exercise (see methods).

Blood pressure measurements were also taken from eight heterozygous horses (HET) at rest.

Horses homozygous for SPH had, on average, significantly higher systolic blood pressure

(SBP), diastolic blood pressure (DBP) and mean arterial pressure (MAP) during exercise,

measured directly after the last uphill interval (Figure 3A-C). In addition, five mins after the

uphill interval, the SPH group showed significantly higher SBP, MAP, and pulse pressure (PP)

relative to the EPH group (Figure 3D-F). Horses homozygous for SPH also had significantly

higher SBP and PP 20 mins after completing the exercise, compared to EPH horses (Figure

3G-H). Twenty mins after the exercise EPH horses were back to resting values, while it took

ten additional mins for SPH horses to recover to resting values. There were no statistically

significant differences in blood pressure measurements at rest, before the exercise.

Furthermore, SPH horses had a higher heart rate than EPH horses during exercise (97.5 vs.

86.4) and five mins after exercise (85.7 vs. 77.0), although the difference was not significant.

All blood pressure values are presented in Supplementary Table S5.

Figure 3. Blood pressure measurements during and after exercise in horses homozygous for the sub-elite performing haplotype (SPH) and horses homozygous for the elite-performing haplotype (EPH). A-C: Blood pressure during exercise measured directly after the uphill interval (SPH n = 5, EPH n=7). D-F: Blood pressure during exercise measured five mins after the uphill interval (SPH n = 5, EPH n=7). G-H. Blood pressure measured 20 mins after completing the exercise (SPH n = 5, EPH n=11). SBP: Systolic blood pressure, DBP: Diastolic blood pressure, MAP: Mean arterial blood pressure, PP: Pulse pressure. PP is defined as the systolic blood pressure minus the diastolic blood pressure.

Plasma concentrations of EDN1 and EDN3 were significantly different between the

haplotype groups

ELISA tests were used to measure the plasma concentration of EDN1 and EDN3 at rest and

during exercise for each haplotype group (Figure 4). Horses homozygous for SPH (n=8) had a

significantly higher plasma concentration of EDN1 both at rest and during exercise compared

to the other groups (Tukey´s HSD test). In addition, SPH horses had a significantly lower

plasma concentration of EDN3 at rest and during exercise (borderline significant) compared to

horses homozygous for the EPH (Figure 4). Individual plasma values are presented in

Supplementary Tables S6 and S7.

Figure 4. Plasma concentrations (pg/ml) of EDN1 and EDN3 at rest and during exercise in horses

homozygous for the sub-elite performing haplotype (SPH) (n=8), horses heterozygous for the EPH

and SPH (HET) (n=11) and horses homozygous for the elite-performing haplotype (EPH) (n=21).

ANOVA and Tukey´s HSD test were performed. Standard deviations are presented as error bars.

Significant differences in protein expression between haplotype groups

To identify potential biological pathways including EDN1 and EDN3, and to compare the expression of different proteins between EPH (n=6) and SPH (n=6) homozygous carriers, we performed a global relative quantitative proteomic analysis using Tandem Mass Tag (TMT) and mass spectrometry (MS), using plasma samples taken at rest or during exercise. In total, 582 distinct proteins were identified, but neither EDN1 nor EDN3 (Supplementary Table S8). Proteins quantified from at least three horses from each haplotype group were included in the statistical calculations. This resulted in 383 proteins that were analyzed in the five following comparisons; 1) exercise vs. rest in SPH, 2) exercise vs. rest in EPH, 3) EPH vs. SPH at rest, 4) EPH vs. SPH during exercise, 5) paired ratio analysis, where the individual difference at rest and exercise was compared between EPH and SPH groups.

Fifty proteins were found to show significantly different expression levels across the five comparisons (P < 0.05; Supplementary Table S8). Table 7 presents the top proteins with the highest fold change in each comparison, including all proteins with different levels that were significant in more than one comparison.

Place for Table 7, see end of document

When comparing protein levels during exercise vs. at rest (comparisons 1 and 2), there were 21 proteins in the SPH group and three proteins in the EPH group with significantly different levels (Supplementary Table S8). A gene ontology (GO) analysis for the 21 proteins showing changing expression profiles at rest vs. during exercise in the SPH group revealed pathways related to blood coagulation, hemostasis, coagulation and regulation of body fluid levels (Figure 5A) (32). When comparing EPH and SPH at rest (comparison 3), there were 23 proteins

upregulated in the SPH group (Supplementary Table S8), implicated in pathways related to

immune response and complement activation (Figure 5B). There were no statistically

significant differences in protein levels when comparing EPH and SPH during exercise

(comparison 4). The paired ratio comparison (comparison 5) revealed 19 proteins showing

significant difference in the individual difference at rest and exercise in SPH and EPH groups

(Supplementary Table S8). These proteins are involved in pathways related to immune

response, complement activation and catalytic and hydrolase activity (Figure 5C).

Figure 5. A. The 20 most significant pathways from the GO analysis based on the significant proteins

in the comparison of exercise vs. rest in the SPH (sub-elite performing haplotype) group. B. All

significant pathways from the GO analysis based on the significant proteins in the comparison of SPH

(sub-elite performing haplotype) and EPH (elite performing haplotype) at rest. C. The 20 most

significant pathways from the GO analysis based on the significant proteins in the paired ratio analysis

of SPH vs. EPH.

Discussion

The EDN3 protein has an established functional role in blood pressure regulation in humans and other species. However, to the best of our knowledge, there are no published reports establishing associations between the *EDN3* gene and equine athletic performance, except for our previous selective sweep study (13). Given the role of EDN3 in blood pressure regulation, we hypothesized that 1) an underlying phenotype for elite athletic performance is blood pressure regulation, and 2) the identified non-coding region harbors a regulatory element that acts on the *EDN3* gene.

All the SNPs defining the haplotypes were found to segregate in the genome of horses that lived prior to the rise and spread of the modern domestic lineage around ~4,200 years ago (DOM2) (31). In fact, temporal allelic trajectories portray an increase in the frequency of the alternative allele, from 7,500-5,500 years ago, prior to the earliest archaeological evidence of horse husbandry (30,33). The concerted rise in frequency of all alternate alleles following the spread of the DOM2 lineage is compatible with a haplotype undergoing positive selection, possibly following artificial selection for animals showing improved athletic performance. Interestingly, the allele trajectories at *EDN3* are in striking contrast with those previously described at the *MSTN* locus, driving performance in short-distance racing (6), which indicated selection starting within approximately the last 1,000 years only. This may indicate performance traits representing recurrent selection targets in the history of horse domestication. We next demonstrated significant differences in blood pressure values, as well as protein levels of EDN1 and EDN3, when contrasting homozygous EPH to SPH CBT carriers. The 5.5 kb region contained multiple marks pointing at a potential *cis*-regulatory module. First, the region spanned an equine enhancer (H3K4me1) active in tissues relevant to racing performance. Second, comparative evidence from human gTEX v8 implicated this region in the regulation

of *EDN3* expression across multiple tissues. Third, of the 14 SNPs forming the haplotypes,

alleles from 12 showed differential transcription factor binding affinities, suggesting

differential regulatory activity between the EPH and SPH carriers. Intriguingly, the SPH

haplotype was predicted to have binding affinities for both immune system actors, NFATC2

and IRF2, where IRF2 has been shown to be a negative regulator of NFAT target genes (34)

(Supplementary Table S4). On the other hand, the EPH haplotype harbored motifs for factors

involved during embryogenesis and cell differentiation, e.g., SRF, HOXA5 and CREB1 (35–

37) (Supplementary Table S4). These differing regulatory roles may point to the multifaceted

role of EDN3, from early neural crest development and proper function of enteric neurons and

melanoblasts (38). The EDN3 ligand has a broad expression range across tissues and has been

implicated in diseases such as Hirschsprung's disease, Multiple Sclerosis and Waardenburg

syndrome in humans, as well as melanocyte development in mice (39–43). However, a key role

for this ligand is in blood flow homeostasis.

Endothelins are involved in one of the most potent vasoregulatory systems, where EDN1 and

EDN3 have opposite effects and act synergistically to regulate blood pressure (18). EDN1

increases blood pressure by vasoconstriction, while EDN3 is involved in nitric oxide release,

which results in vasodilation and decreased blood pressure (18). Additionally, EDN3

stimulates the secretion of vasopressin, which increases the blood volume by retaining water

in the kidneys, while EDN1 has the opposite effect in the kidneys (44–47). In the present study,

horses homozygous for EPH showed significantly higher plasma concentrations of EDN3 and

lower blood pressure during exercise. In contrast, horses homozygous for SPH had a

significantly higher plasma concentration of EDN1 and higher blood pressure during exercise.

Blood pressure depends on cardiac output and resistance in peripheral blood vessels, while

total peripheral resistance is affected by the diameter of the arteries and the viscosity of the

blood. An increase of EDN3 likely leads to a decrease in blood pressure and an increase in

vasodilation of the blood vessels, primarily in the working muscles and the lungs. This means

that blood can flow more easily, allowing improved arterial and venous oxygenation. This

would, in turn, be beneficial for athletic performance. In humans, infusion of EDN3 caused a

decrease in mean arterial blood pressure and vasoconstriction to the visceral organs (48). On

the other hand, infusion of EDN1 leads to an increased mean arterial pressure and decreased

cardiac output (49,50). This is comparable to a study on horses during anesthesia, which

revealed that horses receiving nitric oxide, a vasodilator, had better arterial and venous

oxygenation and lower EDN1 plasma concentration than the control group without nitric oxide

administration (51). In our study, we did not see any effect of exercise on the endothelin levels,

as there were no statistically significant differences in EDN1 or EDN3 levels when contrasting

resting and exercising. This is in concordance with previous studies in which no differences in

EDN1 levels before and during exercise were observed in healthy horses (22,52). In terms of

exercise performance, the present study showed that SPH homozygous horses had a slower

recovery than the horses carrying the EPH. This could be because the SPH group had higher

blood pressure during the exercise, so it took longer for these individuals to return to resting

blood pressure values.

Although the current study demonstrated statistically significant differences in EDN1 and

EDN3 levels between the different haplotype groups, there were no differences in blood

pressure at rest. In adult horses, normal blood pressure at rest is approximately 130/95 mmHg

(systolic/diastolic), in line with the values observed in the current study (53). In humans, it has

been demonstrated that elevated blood pressure at rest is negatively correlated with athletic

performance (21). Individuals with elevated blood pressure had significantly lower maximal

oxygen consumption, ventilatory anaerobic thresholds and heart rate reserves (difference

between an individual's resting heart rate and maximum heart rate) (21). However, unlike

humans, high blood pressure in horses is uncommon and most blood pressure measurements

on horses are performed to evaluate and monitor hypotension. An increase in blood pressure at

rest is most commonly seen as a result of diseases such as laminitis, chronic renal failure or

equine metabolic syndrome (54–56).

The spleen represents one important organ for the distribution of blood, acting as a reservoir

for red blood cells. The equine spleen reservoir is larger than in any other domestic animal

(57). When there is an increased demand, red blood cells stored in the spleen can be released

into the system (57). During exercise, the spleen is contracted, which causes an increase in the

volume of circulating blood cells, while the volume of plasma is unchanged or even reduced.

This leads to an increase in the packed cell volume/hematocrit (i.e., the proportion of blood

made up of red blood cells), hemoglobin concentration, and red blood cell count (57). Previous

studies have demonstrated significantly larger spleen size in horses used for racing compared

to breeds that are not traditionally used for racing (58). In addition, one study demonstrated a

significant correlation between blood pressure and spleen volume (59), with the splenic volume

decreasing when hypertension was induced. However, when hypotension was induced, there

were no significant changes in the splenic volume (59). The results by (59) demonstrated the

importance of the spleen for the cardiovascular system in horses. Further studies examining the

relationship between blood pressure, spleen volume and expression of EDN3 represent

promising avenues for enhancing understanding of the cardiovascular system in horses in

relation to exercising.

Our global proteomic analysis resulted in the identification of a total of 582 proteins. Only one

such protein, Cathepsin X, was encoded as part of the 20 protein-coding genes located within

2 Mb surrounding the genomic region investigated (ECA22: 45,718,095 – 47,718,095). In

contrast to the ELISA analysis, none of the endothelins were detected. A major obstacle in MS-

based plasma proteomics is the large dynamic range in protein concentrations and that the

abundant high molecular weight proteins mask the identification and quantification of lower

molecular weight proteins, such as the endothelins. The highly abundant proteins contribute to

approximately 90 % of the total protein content in plasma. In this study, plasma depletion of

the most abundant proteins was performed to facilitate the quantification of less abundant

proteins and to improve the sequence coverage (60). However, a limitation of the present study

is the quality of the available proteome database for *Equus caballus*, which only includes a

limited number of proteins. Additionally, many proteins were excluded due to a low number

of biological replicates, and in total, 383 proteins could be included in the comparisons between

groups.

When comparing protein levels at rest and during exercise for the different haplotype groups,

we observed a clear difference in the number of regulated proteins (Supplementary Table S8).

In the SPH group, 21 proteins were detected at higher levels at rest, than during exercise. The

GO analysis of these proteins identified several pathways related to injuries, including blood

coagulation, hemostasis, and wound healing (Figure 5A). Several of these proteins also have

functions related to the immune system. When comparing the protein levels between the

haplotype groups at rest, 23 proteins were found regulated. The GO analysis revealed

significant pathways mainly related to the immune system. This is consistent with previous

studies in which an increase of plasma proteins involved in pathways related to inflammation,

coagulation and immune modulation was observed (61,62). A previous genome-wide

association analysis of racing performance in CBTs also found correlations for genes related

to immune response (63). Racing horses perform intense exercise, which induces a physical

stress response to obtain energy for the working muscles. The released stress hormones (e.g.,

cortisol) stimulate gluconeogenesis and lipolysis (64). Additionally, the stress hormones affect

inflammatory and coagulation processes (61,62,65,66). In a non-inflammatory state, intact

endothelial cells maintain blood flow and control coagulation. However, chronic inflammation,

oxidative stress and mechanical stress from turbulent blood flow negatively affect the function

of the endothelial cells (67,68). Furthermore, in a human study, multi-omics changes in

response to exercise were observed, involving pathways related to energy metabolism,

oxidative stress and immune response (69). Potentially, SPH horses have more difficulties

recovering from exercise due to elevated blood pressure, resulting in higher levels of immunity-

related proteins and proteins related to injury healing in the plasma at rest.

It is interesting to note that the protein showing the highest expression difference between SPH

and EPH groups at rest was the proteasome subunit beta (PSMB10). Interestingly, the *psmb10*

gene has previously been linked to blood pressure regulation in mice. Knockout of the *psmb10*

gene was protective for high blood pressure, cardiac fibrosis and inflammation in DOCA salt-

treated mice (70). Although the *PSMB10* gene may be involved in blood pressure regulation in

the horse, it is notable that there were no differences in blood pressure between the haplotype

groups at rest. In the EPH group comparison of protein ratios at rest versus exercise, there were

only three proteins with significant differences. Notably, one of the proteins detected was the

secretoglobin family 1D member (LOC102150542). Secretoglobins are known to have anti-

inflammatory effects in airway disorders such as asthma in both horses and humans (71,72).

The increased levels of this protein may play an anti-inflammatory role during exercise by

preserving the function of the vascular endothelial cells, contributing to a higher tolerance for

intense exercise.

Several studies have demonstrated the effect of the *EDN1* gene and the endothelin receptor

gene, *EDNRB*, in horses and humans (73–81). Studies have suggested EDN1 as a marker for

different diseases in horses, for example, hyperinsulinemia, neonatal diseases, and cardiac

disease (76–78). Moreover, the EDNRB receptor is known to be associated with lethal white

foal syndrome (73–75). However, to the best of our knowledge, there is only one study

performed on horses involving the *EDN3* gene (13). The current study demonstrates significant

differences in blood pressure and plasma levels of both EDN3 and EDN1 in CBTs with

different haplotypes. Although the identified gene variant may also act on other genes in the

region or other parts of the genome, we believe that we have identified a regulatory variant

influencing the transcription of *EDN3*. Based on the ELISA results and racing performance

data, the identified candidate SNP appears to have an additive effect, with homozygous carriers

and heterozygous horses performing significantly better than homozygous wild-type horses.

The obtained results in horses emphasize the importance of the endothelin system as critical

for performance and support the functional role previously shown for EDN3 in human blood

pressure regulation.

Conclusion

Here we demonstrate, for the first time, a significant association between *EDN3*, blood pressure

regulation and athletic performance in horses. We identified a 5.5 kb candidate genomic

interval and showed that horses homozygous for the elite performing haplotype had a higher

plasma concentration of EDN3, lower plasma concentration of EDN1, and lower blood

pressure during exercise compared to horses with the sub-elite performing haplotype. Horses

carrying the elite performing haplotype also recovered from exercise faster than those with the

sub-elite haplotype. The results reported here are important for understanding the biological

mechanisms behind blood pressure regulation, as well as inflammation and coagulation

systems in relation to racing performance in both horses and other mammals, including

humans.

Materials and Methods

Horse material and DNA extraction

A summary of the horses included in the study, and specifying which experiments they were

used for, is presented in Supplementary Table S2. Genomic DNA from all horses was extracted

from hair roots or blood samples. For hair preparation, 186 μL of 5 % Chelex® 100 Resin (Bio-

Rad Laboratories, Hercules, CA, USA) and 14 μL of proteinase K (20 mg/mL; Merck KgaA,

Darmstadt, Germany) were added to each sample. The mixture was incubated at 56°C for two

hours at 600 rpm and inactivated with proteinase K for 10 mins at 95°C. Blood samples were

extracted on the Qiasymphony instrument with the Qiasymphony DSP DNA mini or midi kit

(Qiagen, Hilden, Germany).

Selective sweep racing performance association analysis with additional horse

material

We used the previously identified 19.6 kb selective sweep region, associated with CBT racing

performance, as the starting point for this analysis (ECA22:46,702,297-46,721,892;

EquCab3.0) (13). In that study, g.46717860-T was the variant most significantly associated

with elite performance, and LD to this variant was used to define a five SNP haplotype ($r^2 >$

0.9). We extracted the 19.6 kb region from an existing 670K Axiom Equine Array genotyping

experiment of 661 CBTs (including the 400 horses used for the haplotype analysis in the 2018

study) (13,63). Using g.46717860 as the lead variant, we calculated pairwise r^2 across the

region, and performed racing performance association analyses between this variant, and any

additional variant with $r^2 > 0.6$. Given the high LD, a significance threshold was set at $P \leq 0.05$.

The available phenotypic records included a) pedigree information b) *DMRT3* genotypes and

c) performance data as individual race records for each horse (Swedish Trotting Association

(Svensk Travsport) and the Norwegian Trotting Association (Norsk Rikstoto). The *DMRT3*

variant was genotyped using StepOnePlus Real-Time PCR System (Thermo Fisher Scientific,

Waltham, MA, USA) with the custom TaqMan SNP genotyping assay (82), and included as a

factor in the model, since previous studies have demonstrated major effects of the variant on

harness racing performance (10,12). Only competitive races were included (i.e., premie and

qualification races were excluded). The performance data included: number of starts, number

of wins, number of placings (the number of times a horse finished a race in first, second or

third place), fastest kilometer (km) time in a race where the horse did not gallop (in seconds),

earnings and earnings per start. The earnings for most horses were recorded in Swedish

currency (SEK). Earned prize money in Norwegian currency was converted to SEK based on

the average exchange rate for the year the race took place.

All statistical analyses were performed in the R statistical environment. The performance data

were tested for normality by computing the skewness coefficient using the package moments

v0.14. Non-normally distributed values were transformed, i.e., log transformed values (log10

+1) were used for wins and placings, number of starts were square root transformed and

earnings and earnings per start were transformed according to the previously reported formula

(ln(earnings + 1000)) (83). Genetic variants were tested in a linear model using ANOVA as a

post hoc test. Number of starts, age at first start, sex, birth year, country of birth, and the

DMRT3 genotype were included as fixed effects, when significant. Significant values ($P \leq$

0.05) were further tested using the function lsmeans (Least-Squares Means) with the package

emmeans followed by the multiple comparison test Tukey's HSD-test.

Additional variant discovery

To find and assess variants missing from the Axiom Equine Array, but with the potential to

influence racing performance, we performed both Illumina short-read WGS and MinION

Oxford Nanopore targeted sequencing of PCR products.

First, two CBTs and two SBs were selected for WGS based on their g.46717860 genotype and

their performance values. From each breed, horses homozygous for opposite alleles of SNP

g.46717860 (CC respectively TT) with either high (TT) or low (CC) earnings per start were

selected for sequencing (Supplementary Table S4). Paired-end 150 bp Illumina HiSeqX data

was generated to a depth of 15X. Read data were processed according to the GATK 3.8-0 best

practices (https://software.broadinstitute.org). Briefly, raw data were trimmed with TrimGalore

0.4.4 and quality checked with QualiMap 2.2. The reads were aligned to the EquCab3.0

reference genome (84) (GCF_002863925.1) using BWA-MEM 0.7.17. SAM files were

converted to BAM files using SAMtools, and Picard 2.10.3 was used to sort the BAM files and

remove potential PCR duplications. The BAM files were recalibrated in two steps using

BaseRecalibrator and PrintReads. Variant calling was performed using HaplotypeCaller with

a list of known variants in horses downloaded from the NCBI web page annotation release 103

(https://www.ncbi.nlm.nih.gov). Structural variants were called using default parameters in

FindSV comprising the software programs Manta, TIDDIT and CNVNator

(https://github.com/J35P312/FindSV).

Next, as input for Oxford Nanopore sequencing (ONT, Oxford, UK), a 25 kb PCR product was

designed to span the 19.6 kb sweep region and adjacent repeats. Primers were designed in

Primer3Plus and included unique 10 bp barcodes (Supplementary Table S9). DNA input was

taken from four CBTs with g.46717860-TT genotype and high performance and four CBTs

with the opposite features, including the two CBTs from the WGS analysis (horse selection as

per WGS above and Supplementary Table S2). The PCR region was amplified on a ProFlex

PCR cycler (Thermo Fisher Scientific, Waltham, MA, USA) and the long-range PCR using

AccuPrime™ Taq DNA Polymerase, High Fidelity kit following manufacturers specifications.

The sizes of PCR products were confirmed on 0.8 % TBE Agarose gels, and product

concentration quantified with Agilent 4150 TapeStation using the Agilent Genomic DNA

ScreenTape Assay (Agilent Technologies, Santa Clara, CA, USA). Samples of the respective

PCR products were pooled in equimolar concentration and libraries were prepared using the

Genomic DNA by Ligation (SQK-LSK109) protocol. The samples were loaded and sequenced

on the MinION. Demultiplexed sequencing data were mapped to EquCab3.0 ECA 22

(GCF_002863925) using minimap2 v. 2.4 sequence alignment program

(https://github.com/lh3/minimap2). The resulting SAM files were converted to BAM format

using samtools v. 1.8. Single nucleotide variants were called using the variant calling tool

longshot v. 0.4.3 with default parameters (https://github.com/pjedge/longshot). Structural

genomic variations were identified with NanoSV v. 1.2.4 software

(https://github.com/mroosmalen/nanosv), with default parameters. Only variants with a quality

above 20 were retained.

Genotyping

The largest INDEL identified from the WGS data was analyzed for association with racing

performance traits in 369 CBTs, using linear models. For MassArray SNP genotyping the

number of variant positions available was limited to 25. For this reason, 24 SNPs and one

INDEL, evenly distributed across the 19.6 kb region, were selected for genotyping. In total,

DNA from 412 horses of 13 different breeds and four donkeys were genotyped (Supplementary

Table S10). The horses were carefully selected to include both traditional performance horse

breeds, as well as non-performing breeds. Genotyping was performed at the Mutation Analysis

Facility at Karolinska University Hospital (Huddinge, Sweden), using iPLEX® Gold chemistry

and the MassARRAY® mass spectrometry system (85) (Agena Bioscience, San Diego, CA,

USA). Multiplexed assays were designed using MassARRAY® Assay Design Suite v2.2

software (Agena Bioscience, San Diego, CA, USA), genotyping the 25 markers in one reaction

per sample. The protocol for allele-specific base extension was performed according to Agena

Bioscience's recommendation. Analytes were spotted onto a 384-element SpectroCHIP II

array (Agena Bioscience, San Diego, CA, USA), using Nanodispenser RS1000 and

subsequently analyzed by MALDI-TOF on a MassARRAY® Analyzer 4 mass spectrometer

(Agena Bioscience, San Diego, CA, USA). Genotype calls were manually checked by two

persons individually using MassARRAY® TYPER v4.0 Software (Agena Bioscience, San

Diego, CA, USA).

The variant g.46717860 was not included as part of the MassArray, but was genotyped for 170

horses (9 Ardennes, 96 CBTs, 4 donkeys, 4 Finnhorses, 2 Fjord horses, 2 Gotlandsruss, 10

North-Swedish draught horses, 3 Pzrewalski and 40 SBs) using the StepOnePlus Real-Time

PCR System (Thermo Fisher Scientific, Waltham, MA, USA) and the custom TaqMan SNP

genotyping assay (82). The reaction volume of 15 µl comprised 1.5 µl DNA, 0.38 µl

Genotyping Assay 40X, 7.5 µl Genotyping Master Mix 2X, and 5.62 µl deionized water.

Haplotype analysis

Pairwise r^2 calculations with all 24 SNPs and g.46717860 were performed using the LD-

function in R. Based on the high LD between all genotyped SNPs, it was possible to impute

g.46717860 alleles for ungenotyped samples. All SNPs with a pairwise r^2 value >0.6, plus two

flanking SNPs were selected for haplotype analysis. Haplotype racing performance association

tests used a GLM regression analysis, which was performed in 180 CBTs and 38 SBs using the

haplo.glm function from the haplo.stats package in R (86). The model included the effects of

sex, age, country of registration, *DMRT3* genotype (only for the CBTs) and number of starts,

when significant. All CBTs had been genotyped for the DMRT3 mutation, using the

StepOnePlus Real-Time PCR System (Thermo Fisher Scientific, Waltham, MA, USA) with

the custom TaqMan SNP genotyping assay (82). Haplotypes with a frequency < 2 % were

considered rare, and were not analyzed for association with performance. A minimum shared

haplotype of 14 SNPs between CBTs and SBs was used to define the elite-performing

haplotype (EPH) and the sub-elite performing haplotype (SPH).

Association analysis with athletic performance

Allele calls at SNP g.46718095 were a proxy for elite-performing haplotype (EPH) and the

sub-elite performing haplotype (SPH). Samples not genotyped on the MassArray were

genotyped using the StepOnePlus Real-Time PCR System (Thermo Fisher Scientific,

Waltham, MA, USA) with the custom TaqMan SNP genotyping assay (82). Association with

racing performance was performed as per "Selective sweep racing performance association

analysis with additional horse material" using 539 CBTs born between 1994 and 2017 and 271

SBs born between 2005 and 2014. Horses of 13 different breeds and three other types of equids

(wild ass, onager and donkey) were also genotyped to investigate the distribution of allele

frequencies.

Functional bioinformatic analyses

ChIP-seq data were provided by the Equine Functional Annotation of Animal Genomes

(FAANG) team (www.faang.org) (24,25) including four histone marks (H3K4me1, H3K4me3,

H3K27ac and H3K27me) across nine tissues (adipose, brain, heart, lamina, liver, lung, muscle,

ovary and skin) from two Thoroughbreds. FAANG bam and bed files from the different

available tissues, the four histone marks and the input DNA control (project: PRJEB35307)

were downloaded to a local server. First, the integrity of the data was checked using checksums.

Bam files were sorted with samtools and bedgraphs were subsequently created using bedtools

(87,88). The Integrative Genomics Viewer (IGV) browser was used to visualize the output data

along with the EquCab3.0 fasta file and its gff3 annotation file (84,89). The presence of

H3K27ac and H3K4me1 was considered to mark active enhancers, whereas H3K4me1 and

H3K27me3 marked poised enhancers. H3K27ac and H3K4me3 marked active promoters and

H3K27me3 was considered to mark repressed regions. The region defined by the minimum

shared haplotype (see above) was investigated using IGV for the different histone marks in all

available tissues.

A comparative analysis to hg38 was conducted for the minimum shared haplotype region. Co-

ordinates were lifted in UCSC (liftOver) and potential overlaps between ENCODE Candidate

Cis-Regulatory Elements (cCREs), ENCODE Integrated Regulation tracks and ORegAnno

Regulatory elements were annotated (90,91). For each of the SNPs in the minimum shared

haplotype region, the potential of alleles to alter transcription factor binding potential was

tested with sTRAP (92). Here, 31 bp sequences centered on the allele of interest were tested

using JASPAR vertebrates, on a background model of human promoters. In the cases where

two SNPs were within 10 bp of each other, they were tested as a single phased fragment.

Ancient DNA screening

We leveraged the availability of an extensive ancient genome time-series in the horse to assess

the temporal trajectory and spatial distribution of the SNPs in the minimum shared haplotype

within the CBTs, in the past. More specifically, we re-aligned against the EquCab3.0 reference

genome (84) supplemented with the Y-chromosomal contigs from (93) and a sub-selection of

the sequencing data from (2,27,29,30,94,95), representing a total of 431 ancient genomes. Alignment files were generated using the Paleomix pipeline (96), and further rescaled and trimmed, following the procedures described in (31). This procedure ensured minimal sequencing error rates, especially at sites potentially affected by post-mortem DNA damage. The number of occurrences of each individual allele was counted for each individual using ANGSD (v0.933), with the -baq 0, -remove_bads -minMapQ 25 -minQ 30 -rmTriallelic 1e-4 -SNP_pval 1 -C 50 parameters (97). The resulting occurrences were tabulated together with metadata available for each individual, consisting of their corresponding archaeological site (name and GPS coordinates) and their average calibrated radiocarbon date (or age as assessed from the archaeological context otherwise). The tabulated file was visualized using the mapDATAge package (26), which provides spatial distributions within user-defined time periods and/or spatial ranges and estimates allelic trajectories (i.e., allelic frequencies through time) based on the resampling of individual allelic counts. The final allelic trajectories were assessed considering time bins of 2,000 years (step-size = 500). To explore the genetic variation for the SNPs within the haplotype, we considered the 14 polymorphic sites i.e., ECA22:46703380, 46708110, 46708983, 46713478, 46715974, 46717451, 46717528, 46717742, 46717854, 46717999, 46718095, 46718113, 46718361, 46718436, 46718895, 46718964, 46719042, 46719906, 46720030, 46720999, and 46721822.

Exercise test, sample collection and blood pressure measurements

Horse material

Thirty horses were enrolled in the exercise test (Supplementary Table S2). All horses were genotyped for SNP g.46717860 and g.46718095 using the StepOnePlus Real-Time PCR System (Thermo Fisher Scientific, Waltham, MA, USA) with the custom TaqMan SNP genotyping assay as above (82). The two SNPs were used as a proxy for the haplotype variants

homozygote EPH (TT respectively CC) or homozygote SPH (CC respectively TT). Horses

were then divided into three groups based on haplotype variant, i.e., homozygote EPH or SPH

and HET (heterozygote). The horses were carefully selected to include an even sex and age

distribution across the different groups. The age of the horses ranged from two to 13 years

(mean age = 5.5 years). All horses, except for the two-year old, were in competing condition.

However, due to various reasons including practical problems, blood pressure measurements

were not taken from all horses at all time points and some horses only had measurements at

rest (Supplementary Table S2). For the ELISA and proteomics analysis, plasma samples were

collected from 40 horses, before and/or during exercise. As for blood pressure measurement,

some horses only had samples collected at rest.

Exercise

All horses, except for the two two-year old horses, were challenged on the same standardized

exercise session (n=28). The test was performed on several days during autumn and winter of

the years 2020 and 2021. Each horse, with a trainer, completed a 1.5-hour exercise route that

included both flat and uphill interval training. This route was commonly used in their daily

training program. A Polar M460 sensor with GPS (Polar Electro, Kempele, Finland) was used

to record heart rate (HR), speed, distance and amplitude during exercise. The horses performed

the exercise on a track with hilly terrain in the forest. The uphill interval section consisted of

the horses trotting four times up a hill with four degrees inclination. Each uphill interval lasted

for five mins with a distance of 1.7 km. The HR for each horse reached above 200 beats/min.

After the uphill intervals, the horses walked down to the flat part of the track, where they

completed 10 mins of intervals, at a speed of 35 km/h and HR of about 180-200 beats/min.

Finally, the horses jogged one km back to the stable at a speed of 15-20 km/h. During the

exercise, the average HR reached above 190 beats/min for approximately eight mins. Due to

their young age the two-year old horses (n=2) performed a standardized exercise session at the

racetrack. Following warm-up, they performed a number of intervals with HR reaching over

200 beats/min. The two-year old horses performed the exercise test at the same time.

Blood pressure measurements

The following measurements were collected with a Cardell Veterinary Monitor 9402 cuff

(Cardell 10), at the middle coccygeal artery: systolic, diastolic, mean arterial pressure, pulse

pressure, and heart rate. The same cuff size was used for all the horses. The monitor has

previously been successfully evaluated for use in horses (98). Blood pressure was measured at

rest in the stable before the exercise and during exercise (directly and five mins after the last

uphill interval) (Figure 6). Additionally, blood pressure was measured every ten mins after

exercise, up to 40 mins after the exercise was completed. In order for the horse to acclimate to

the equipment, the Polar M460 sensor and the cuff were put on at rest, before the exercise route,

and the horse was left in the box for 10 mins before measurements were taken. Similarly, blood

pressure was measured five times at rest for the horse to get used to the measurement procedure.

The measurements with the highest and lowest systolic blood pressure values were removed.

During and after exercise the blood pressure was measured three times and an average

calculated. For the association analyses with blood pressure, the mean values of the triplicates

were analyzed using ANOVA and Tukey's HSD test.

Figure 6. A schematic overview of the parameters recorded during the exercise for each horse,

including heart rate, speed and altitude. The arrows indicate when blood samples and blood pressure

measurements were taken.

Biological sample collection and analysis

Blood samples were collected at rest before the exercise and during exercise directly after the

last uphill interval. Blood was also collected for plasma extraction. Here, a sample was

collected in an EDTA tube and mixed gently four times after sampling, to allow the EDTA to

mix with the blood, then centrifuged at 1000 x g (3000 rpm) for 20 mins as soon as possible

and within two hours after sampling. The tubes were kept cold until centrifugation. The EDTA

plasma was aliquoted in Eppendorf tubes and immediately frozen at -20°C and within five days

put at -80°C until analysis.

ELISA analysis of EDN1 and EDN3

Plasma EDN1 concentrations were measured using the Horse Endothelin ELISA Kit

PicoKine™ EK0945-EQ (Boster Bio, Pleasanton, CA, USA). Plasma EDN3 was quantified

with the horse EDN3 ELISA Kit MBS9901200 (MyBioSource, San Diego, CA, USA). The

analyses were performed according to the manufacturer's manual. A standard curve was

created, and the samples were run in duplicates. Mean values of the concentration (pg/ml) were

used in the statistical analysis, and the standard deviation (SD) and coefficient of variation

(CV) were calculated. Differences in concentrations between the three groups (EPH, HET and

SPH) before and during exercise were analyzed using ANOVA and Tukey´s HSD test.

Significance was set at P-values ≤ 0.05.

Relative quantitative proteomic analysis

A global relative quantitative proteomic analysis was performed to identify differentially

expressed proteins in plasma before and during exercise. Six horses of each haplotype group

(EPH or SPH homozygote) were included in the analyses (Supplementary Table S2). The

horses were selected based on earnings per start, age and sex. The two-year-old horses were

matched, one from each haplotype group, as they did not perform the same standardized

exercise as the other horses.

Sample preparation

The plasma samples and references (10 µl) were depleted from albumin and IgG, using the

Pierce™ Albumin/IgG Removal Kit (Thermo Fisher Scientific, Waltham, MA, USA)

according to the manufacturer's instructions. The reference was a representative pool that

included four groups: EPH at rest, EPH during exercise, SPH at rest and SPH during exercise.

The samples were processed using the modified filter-aided sample preparation (FASP) method

(99). Half the volume, corresponding to approximately 30 µg, was concentrated on Microcon-

30kDa Centrifugal Filter Units (Merck & Co, Kenilworth, NJ, USA), reduced with 100 mM

dithiothreitol at room temperature for two hours, washed several times with 8 M urea and once

with digestion buffer (DB) (25 mM TEAB, 0.5 % sodium deoxycholate) before alkylation with

871 10 mM methyl methanethiosulfonate in digestion buffer for 30 mins. Digestions were

872 performed by adding 1 μg Pierce MS grade Trypsin (Thermo Fisher Scientific, Waltham, MA,

873 USA) in DB and incubating overnight at 37°C. An additional portion of 1 μg trypsin was added,

874 and the samples were incubated for another three hours. Peptides were collected by

875 centrifugation and labeled using Tandem Mass Tag (TMTpro 16plex) reagents (Thermo Fisher

876 Scientific, Waltham, MA, USA), according to the manufacturer's instructions. The labeled

877 samples were combined into two sets, where both sets included all four groups and a reference.

878 The paired samples were analyzed in the same set. Sodium deoxycholate was removed by

879 acidification with 10 % TFA and the sets were desalted using Pierce peptide desalting spin

880 columns (Thermo Fisher Scientific, Waltham, MA, USA) according to the manufacturer's

881 instructions. The two sets were pre-fractionated into 40 fractions with basic reversed-phase

882 chromatography, using a Dionex Ultimate 3000 UPLC system (Thermo Fisher Scientific,

883 Waltham, MA, USA), an XBridge BEH C18 column (3.5 μm, 3.0x150 mm, Waters Corp,

884 Milford, MA, USA) and a gradient from 3 % to 90 % acetonitrile in 10 mM ammonium formate

885 with pH 10.0 for 25 mins. The fractions were concatenated into 20 fractions, dried, and

886 reconstituted in 3 % acetonitrile and 0.2 % formic acid.

887 **Liquid chromatography–Mass Spectrometry/ Mass Spectrometry analysis**

888 The fractions were analyzed on an Orbitrap Fusion Lumos Tribrid mass spectrometer

889 interfaced with an Easy-nLC 1200 liquid chromatography system (Thermo Fisher Scientific,

890 Waltham, MA, USA). Peptides were trapped on an Acclaim Pepmap 100 C18 trap column (100

891 μm x 2 cm, particle size 5 μm, Thermo Fisher Scientific, Waltham, MA, USA) and separated

892 on an analytical column (75 μm x 35 cm) packed in-house with Reprosil-Pur C18 material

893 (particle size 3 μm, Dr. Maisch, Ammerbuch, Germany) at a flow of 300 nL/min using a

894 gradient from 4 % to 80 % acetonitrile in 0.2 % Formic Acid over 90 mins. The Orbitrap was

895 equipped with the FAIMS Pro ion mobility system alternating between the compensation

voltages of -50 and -75. MS scans were performed at 120,000 resolutions, and the 12 most

abundant doubly or multiply charged precursors were isolated using 0.7 m/z isolation window

and dynamic exclusion within 10 ppm for 60 seconds. The precursors were fragmented by

collision induced dissociation (35 %) and detected in the ion trap, followed by multinotch

(simultaneous) isolation of the top 10 MS2 fragment ions within the m/z range 400-1400, MS3

fragmentation by higher-energy collision dissociation (55 %) and detection of the reporter ions

in the Orbitrap at 50 000 resolution, m/z range 100-500.

Proteomic data analysis

The data files in each set were merged for identification and relative quantification using

Proteome Discoverer version 2.4 (Thermo Fisher Scientific, Waltham, MA, USA). The Uniprot

Equus caballus proteome database (July 2021, combined SwissProt and TrEMBL 44,488

entries) and Mascot search engine v. 2.5.1 (Matrix Science, London, UK) were used in the

database matching with peptide and fragment ion tolerance of 5 ppm and 0.6 Da. Tryptic

peptides with zero missed cleavage, variable modifications of methionine oxidation and fixed

cysteine alkylation were accepted. ProTMT modifications of N-terminal and lysine were

selected. Percolator was used for the peptide-spectrum match validation with a strict false

discovery rate (FDR) threshold of 1 %. The TMT reporter ions were identified with 3 mmu

mass tolerance and no normalization was applied to the samples. For the protein quantification,

only unique peptide sequences at a strict FDR threshold of 1 % with a minimum synchronous

precursor selection match of 65 % and an average S/N above 10, were taken into account. The

reference in each set was used to calculate protein ratios and quantified proteins were filtered

at 5 % FDR.

Processing of quantitative proteome data

The protein ratios of quantified and normalized proteins were processed in Microsoft Excel to

calculate the group means and fold changes. All plasma samples were grouped into two sets;

before and after exercise. Only proteins quantified in at least three horses from each haplotype

group (EPH and SPH), i.e., proteins with at least three biological replicates in each haplotype

group, were included in the analyses. Perseus version 1.6.15.20 was used to analyze differences

between the groups, by applying t-test on log2 protein ratios. Five different comparisons were

made; SPH and EPH at rest and during exercise as well as rest versus exercise within each

haplotype. Additionally, paired ratio analysis was performed, to compare differences in protein

levels during exercise vs. at rest for each haplotype i.e., (SPH-rest/SPH-exercise)/(EPH-

rest/EPH-exercise). For each comparison, fold change values were calculated by dividing the

average values for the compared groups and P-values ≤ 0.05 were considered to be significant.

GO-annotations downloaded from the Uniprot database into Perseus were used to support the

identification of changing biological processes. Furthermore, the GO enrichment analysis and

visualization tool g:GOst from g:Profiler was used to identify the biological processes affected

in each of the different contrasts (32).

Declarations

Ethics approval and consent to participate

Blood sample collection was approved by the ethics committee for animal experiments in

Uppsala, Sweden (number: 5.8.18-15453/2017 and 5.8.18-01654/2020).

Consent for publication

Not applicable.

Availability of data and materials

The datasets used and analyzed during the current study will be deposited to the SRA.

Competing interests

The authors declare that they have no competing interests.

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Authors´ contributions

G.L. designed the project and led the genetic mapping together with J.R.S.M. M.K.R. was supervised with main input from G.L. and R.N. to perform the sampling, genetic mapping and blood pressure measurements. M.Å. designed genotyping assays. A.J. performed genotyping. M.K.R. and K.F. performed haplotype analysis. L.O. performed the mapDATAge analysis. R.N. designed the long-read sequencing experiment, performed ChIP-sequencing and proteomics data analysis. A.T. performed proteomic experiments and analysis. B.D.V. contributed with supervision of M.K.R. and A.R. contributed with mentoring of K.F. B.E .advised the blood pressure measurements. C.M.M. assisted with human comparisons and in the interpretation of the blood pressure results. G.A. performed transcription factor binding site analyses. J.R.S.M. assisted with the comparative genomic analysis and performed the transcription factor binding site analysis. K.F. lead the writing of the manuscript and created figures, with input from all co-authors.

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Table 7. Fold-change and P-value for the top most upregulated proteins in each comparison in the quantitative proteomic analysis, including all

proteins significant for more than one comparison

| | | | | | Comparison | | | | | | | |
| | | | | | 1. SPH exercise vs. rest | | 2. EPH exercise vs. rest | | 3. EPH vs. SPH rest | | 5. EPH vs. SPH Paired ratio | |
Protein function		Protein	Gene	Accession	FC	P-value	FC	P-value	FC	P-value	FC	P-value
Immune System	*Adaptive immune system	Proteasome subunit beta	*PSMB10*	F6SZ35					6.03	0.03	8.2	0.03
	*Adaptive immune system, angiogenesis	CD34 molecule	*CD34*	F6S1S0	4.33	0.03			3.42	0.03	17.97	0.04
	*Antigen binding	Ig-like domain-containing protein	*Unknown*	A0A3Q2HY39					5.58	0.02		
	*Antigen binding	Ig-like domain-containing protein	*Unknown*	A0A3Q2GXU9					5.44	0.03		

*Complement activation, complement and coagulation cascades	Complement C1r	*C1R*	F6Z5L1	4.69	0.02			4.09	0.03
*Complement activation, inflammatory response	C4a, anaphylatoxin	*LOC100059239*	F6XSF7			2.08	0.04	5.6	0.04
*Complement activation, inflammatory response	C3-beta-c	*LOC100060505*	A0A3Q2HWQ6	2.54	0.05			5.83	0.05
*Complement and coagulation cascades, metabolism of proteins	Carboxypeptidase B2	*CPB2*	F6PSH3	4.31	0.01			7.94	0.03

Category	Function	Protein	Gene	Accession								
	*Innate immune system	Cystatin C	CST3	A0A3Q2I5T2	2.32	0.02			4.23	0.007	5.91	0.02
	*Innate immune system	Canopy FGF signaling regulator 3	CNPY3	F7CNK7	5.05	0.04						
	*Innate immune system, neutrophil degranulation, angiogenesis	Serine peptidase inhibitor, Kunitz type 1	SPINK5	A0A5F5Q0T5	3.22	0.04			4.65	0.03	18.49	0.04
	*Small molecule binding, allergenic capacity	Equ c 2 allergen	LOC106781327	A0A3G1Z1K7			0.28	0.001				
	*Small molecule binding, allergenic capacity	Major allergen Equ c 1	LOC100056556	A0A5F5PQP7			0.24	0.04				
Metabolism	*Metabolism gluconeogenesis	Malate dehydrogenase 1	MDH1	A0A5F5PWG8	4.11	0.05						

	*Metabolism of proteins	Tripeptidyl aminopeptidase	TPP1	F6TJX5					2.88	0.04	10.13	0.05
Neuronal	*Neuronal system	Neurexin 2	NRXN2	F6YED9					5.85	0.005		
Others	*Serine-type endopeptidase inhibitor activity	Serine peptidase inhibitor Kazal type 5	SPINT1	A0A3Q2IKF1							6.35	0.04
	*Amino acid metabolism	T1-TrpRS	WARS1	A0A3Q2KTL8	3.51	0.02			3.48	0.01	13.11	0.005
	*Binding of androgens and other steroids	Secretoglobin family 1D member	LOC102150542	A0A3Q2HLK8			0.17	0.0009				
	*Hemostasis and regulation of insulin-like growth factor (IGF) transport and uptake	Fibrinogen gamma chain	FGG	A0A5F5PPB8	2.8	0.05					4.28	0.05

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