DNA methylation patterns of behavior-related gene promoter regions dissect the gray wolf from domestic dog breeds

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Giulia Cimarelli, ^{2,3,4,} ^{7,8} Affiliations "7" and "8" are actually a part of affiliation "2" in the present case, so data from "7" and "8" should be merged into affiliation "2".

The reason is that authors G Cimarelli and Z Viranyi are actually affiliated to the Messerli Research Institute (in addition to the Wolf Science Center). However, Messerli Institute is in close partnership with both the University of Veterinary Medicine, Vienna, the Medical University of Vienna and the University of Vienna; thus in all cases when the Messerli Institute is indicated as an affiliation, these three universities should be indicated as well. As the Messerli Institute is physically located within premises of the University of Veterinary Medicine, Vienna, address of the latter is indicated (Veterinärplatz 1, 1210 Vienna, Austria).

Thus, the correct text for affiliation "2" is: Comparative Cognition, Messerli Research Institute, University of Veterinary Medicine, Vienna, Medical University of Vienna, University of Vienna, Veterinärplatz 1, 1210 Vienna, Austria.

Accordingly, affiliations "7" and "8" should be deleted.

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Abstract

A growing body of evidence highlights the relationship between epigenetics, especially DNA methylation, and population divergence as well as speciation. However, little is known about how general the phenomenon of epigenetics-wise separation of different populations is, or whether population assignment is, possible based on solely epigenetic marks. In the present study, we compared DNA methylation profiles between four different canine populations: three domestic dog breeds and their ancestor the gray wolf. Altogether, 79 CpG sites constituting the 65 so-called CpG units located in the promoter regions of genes affecting behavioral and temperamental traits (*COMT, HTR1A, MAOA, OXTR, SLC6A4, TPH1, WFS1*) —regions putatively targeted during domestication and breed selection. Methylation status of buccal cells was assessed using EpiTYPER technology. Significant inter-population methylation differences were found in 52.3% of all CpG units investigated. DNA methylation profile-based hierarchical cluster analysis indicated an unambiguous segregation of wolf from domestic dog. In addition, one of the three dog breeds (Golden Retriever) investigated also formed a separate, autonomous group. The findings support that population segregation is interrelated with shifts in DNA methylation patterns, at least in putative selection target regions, and also imply that epigenetic profiles could provide a sufficient basis for population assignment of individuals.

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Keywords

DNA methylation Canine Population assignment Domestication Behavior Promoter

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Electronic supplementary material

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Introduction

Epigenetic factors represent a dynamic connection between heritage and environment, making them an ideal prime target for evolutionary processes. Epigenetic states can be highly sensitive to environmental effects (Dolinoy 2008; Turner 2009; Feil and Fraga 2011), but can also be reliably transmitted across generations (Xing et al. 2007; Daxinger and Whitelaw 2010; Gaydos et al. 2014; Bale 2015). It has been theorized that speciation itself begins with variation in epigenetic patterns, followed only later by divergence in DNA sequence (Skinner et al. 2015; Smith et al. 2016). Epigenetic alterations can ultimately induce changes in nucleotide order (Molaro et al. 2011), while on the other hand small-scale polymorphisms apparently contribute to modifications in epigenetic marks and, consequently, transcriptional regulation (Fukuda et al. 2013).

Genome-wide DNA methylation data both on closely related species (human and ape) (Mendizabal et al. 2016) and on wild and domesticated strains of the same species (chicken) (Natt et al. 2012) imply that separate populations might show well distinct epigenetic patterns, at least in selective sweep regions. Besides, evidence indicates that such population-specific DNA methylation differences are already present in the germline (Molaro et al. 2011) and that germline DNA methylation states strongly correspond to the somatic ones (Martin et al. 2011), suggesting that inherited epigenetic profiles both constrain the somatic epigenetic landscape and also distinguish closely related species and strains. However, inter-population DNA methylation comparison studies so far have mainly been conducted on a very limited number of subjects, raising the questions whether the observed differences are generally characteristic to entire populations, and if DNA methylation data could solely be sufficient for successful population assignment of individuals.

As implicated by former inter-population DNA methylation studies, population-characteristic differences can principally be expected to be observed in regions under strong selection pressure, be that natural or artificial. Genes affecting temperamental and cognitive traits apparently always constitute a considerable proportion of such genomic loci (Molaro et al. 2011; Natt et al. 2012; Mendizabal et al. 2016), which is probably expected given that adequate adaptation to any novel environmental challenges generally requires alterations in behavioral functions as well. In fact, behavioral isolation itself is considered as a major driver of population divergence and speciation (Verzijden et al. 2012; Sommer-Trembo et al. 2016).

Domestication and breeding process offers an unrivaled opportunity to investigate selection-induced molecular biological changes, including putative shift in epigenetic patterns (Jensen 2015), and of all domesticated species, it is possibly the dog whose investigation is most intriguing. Artificial selection has created a spectacular morphological and behavioral diversity in the dog (Careau et al. 2010; McGreevy et al. 2013), which has been turned from its large predator ancestor into a popular pet serving man by a variety of ways, and in modern times often sleeping not only in his house but right in his bed. Yet dog and wolf are up today of the same species according to the biological species definition (group of organisms consisting of individuals capable of reproducing fertile offspring). Modern breeding practices have also established multiple closed reproductive populations called breeds, all of which have been selected for a specific desired phenotype—personality traits included (Parker 2012; Wayne and vonHoldt 2012).

In the present study, we compare buccal DNA methylation profiles between promoter regions of behavior-related genes in the gray wolf and three domestic dog breeds selected for characteristically different personality traits: Border Collie, Golden Retriever, and Siberian Husky. Protein products of the investigated genes play part in diverse biochemical pathways and neurotransmission systems (including different monoaminergic and the oxytocin systems as well as endoplasmic reticulum stress signaling) as enzymes, receptors, or transporters. All genes analyzed have variants confirmedly influencing temperament and/or cognitive functions (Shih et al. 1999; Weinshilboum et al. 1999; Gainetdinov and Caron 2003; Nakamura et al. 2006; Lesch 2007; Kato et al. 2008; Neumann 2008; Ptacek et al. 2011), rendering them ideal selection target candidates during domestication and breeding processes.

Materials and methods

Animals studied

DNA methylation analysis was carried out on three dog breeds (Golden Retriever, Siberian Husky, and Border Collie) as well as North American timber wolves. All four populations contained 8 individual animals each. Only male animals were investigated. Mean age \pm SD was 3.4 ± 2.4 , 4.5 ± 2.8 , 3.8 ± 2.1 , and 4.0 ± 1.8 years for Golden Retriever, Siberian Husky, Border Collie, and wolf, respectively. Differences in mean and variance of age were identified as non-significant by one-way ANOVA and Bartlett's test for equal variances (p = 0.8041 and p =

0.6735, respectively). All owners volunteered to participate and gave consent to the genetic analysis of their animals and the use of results for research purposes. Dogs were kept as pets; wolves were a zoo-based population living in packs and in regular physical contact with humans at the Wolf Science Center (WSC), Ernstbrunn, Austria. All wolves were born in captivity, hand-raised at the WSC (from age 10 days to about 5 months) and motivation-based clicker trained as adults. More information on wolf-keeping conditions and dog-wolf comparative ethological research at the WSC can be found on http://www.wolfscience.at. Two of the wolves were cousins, otherwise none of them were closely related. Wolves lived in four physically separated packs. Three of them were alpha wolves. None of the dogs of the same breed were closely related (first- and second-degree relatives were automatically excluded). All Golden Retrievers and all Border Collies lived at different households. Of Siberian Huskies, two animals shared a household in three cases. This study required no special permission of the University Institutional Animal Care and Use Committee (IACUC) as non-invasive animal studies are not regarded as animal experiments by the operative law of Hungary (Act XXVIII of 1998 on the protection and welfare of animals).

Sample collection

Buccal samples were collected from the inner cheek using sterile cotton-tipped swabs. DNA was obtained by a traditional, salting-out procedure (Boor et al. 2002). Briefly, following overnight incubation at 56 °C in 450 μ l cell lysis buffer (0.2 g/l Proteinase K, 0.1 M NaCl, 0.5% SDS, 0.01 M Tris buffer, pH 8.0), samples were RNase treated at room temperature and protein precipitated with saturated (6 M) NaCl. After isopropanol precipitation and ethanol purification, pellets were resuspended in 30 μ l standard Tris–EDTA solution (0.01 M Tris, \pm 0.001 M EDTA, pH 8.0). DNA was kept at –20 °C until bisulfite conversion.

Target region selection

Genes of interest were those with a proven influence on neurobiological processes and behavior in mammals. Only CpG island shores in promoter regions were included (up to 2600 bp upstream from

annotated transcription start site). For gene annotation purposes, genome assembly CanFam3.1 was used (Aken et al. 2016). Prerequisites for target gene selection included (1) canine promoter sequences available in public databases and (2) confirmed relationship between polymorphisms (preferably regulatory region polymorphisms) and behavioral differences in mammals and, if possible, also in dogs. Care was taken that genes affecting distinct signaling pathways and encoding different types of proteins (e.g., with receptor, transporter, or enzymatic activity) were included. The final gene set was randomly selected from a larger set meeting all the selection criteria.

Localization of CpG islands was determined using an in-house MS-DOS application based on the CpG island definition of a region of \geq 200 bp length with a >50% GC percentage and a >60% observed-toexpected (O/E) CpG ratio. Short amplicons covering the highest possible number of CpGs within promoter CpG island shores were preferred. Sequence match of selected amplicons between domestic dog and wolf (as well as golden jackal) was principally checked using a private database (Department of Ecology and Evolutionary Biology, University of California, LA) with the kind permission of Robert K. Wayne. Existence of polymorphisms was also checked in Ensembl, dbSNP, and DoGSD public databases (Sherry et al. 2001; Bai et al. 2015; Aken et al. 2017). Amplicons containing a possibly polymorphic CpG site were excluded from the study.

On the basis of the above considerations, 12 amplicons in 7 genes of interest were selected. Six of those (*COMT, HTR1A, MAOA, OXTR, SLC6A4*, and *TPH1*) encode proteins important in neurosignaling, while *WFS1* encodes an endoplasmic reticulum transmembrane protein. Level of sequence conservation between all canine sequences investigated was at least 98% for the target regions.

DNA methylation analysis

1000 ng genomic DNA quantified by a NanoDrop ND-1000 Spectrophotometer (Thermo Scientific, Wilmington, DE) was bisulfite converted using the EZ DNA Methylation-Gold[™] Kit (Zymo Research, Irvine, CA) according to the manufacturer's protocol. PCR primers were designed by the EpiDesigner online tool (http://www.epidesigner.com). Primer adherence sites are indicated in Supplementary Table 1. Reaction mixture for PCR contained 0.5 units EpiMark Hot Start Taq DNA Polymerase (New England Biolabs, Ipswich, MA), $1 \times$ EpiMark Hot Start Taq Reaction Buffer, 0.2 mM dNTP, 0.2–0.2 μ M of forward and reverse primers, and about 15–20 ng bisulfite-converted DNA template. Amplification was carried out using 45 cycles. Denaturation step took place at 95 °C/30 s and elongation at 68 °C/45 s. For annealing times and temperatures, see Supplementary Table 1. After PCR fragment quality check by gel electrophoresis, amplicons were sent to Agena Bioscience GmbH (Hamburg, Germany) for in vitro RNA transcription by T7 Polymerase, base-specific RNA cleavage and mass spectrometry-based DNA methylation analysis (Ehrich et al. 2005) using MassARRAY[®] EpiTYPER[®] technology (Agena Bioscience, San Diego, CA).

For array validation, percent methylation of a CpG site (OXTR_17_3) showing significant inter-population methylation differences was assessed also by pyrosequencing. Primer sequences were as follows: 5' AGG GTG ATG AAG TTG TAA AAG T 3' (forward), 5' ACA TTT CAT CTT CCT TTA ACA TCA TAT A 3' (reverse), and 5' GGT TTT TTT TTT TTT TGG TTT AGA A 3' (sequencing). Reverse primer was biotinylated on the 5' end. PCR reaction mixture composition was the same as used at amplification for MassARRAY[®] EpiTYPER[®] analysis. Cycling conditions were as follows: Step 1: (95 °C/1 min)/1 cycle; Step 2: (95 °C/30 s, 58 °C/1 min, 68 °C/45 s)/45 cycles; Step 3: (68 ° C/5 min)/1 cycle; Step 4: 8 °C hold. Pyrosequencing was performed in triplicate on a PyroMark Q24 platform using PyroMark Gold Q24 Reagents (Qiagen NV, Venlo, NL).

In silico analyses

Heat map of methylation level per CpG unit was plotted using MultiExperiment Viewer (MeV) version 4.9.0 (http://www.tm4.org). Statistical calculations were performed by GraphPad Prism 5.03 for Windows (GraphPad Software, San Diego, California; http://www.graphpad.com). One-way ANOVA followed by Tukey –Kramer post hoc test was used for assessing differences in percent methylation between populations. Hierarchical clustering dendrograms were obtained by XLSTAT 2015.1 (http://www.xlstat.com) applying Ward's method of minimum variance (with Euclidean distance matrix).

Results

Data quality and study design

Altogether, 93 CpG sites were covered by the amplicons, 79 of which were analyzable by the MassARRAY[®] technology. These comprised the 65 so-called CpG units, which are the smallest possible cleaved fragments analyzed by MassARRAY[®] assay. Where a CpG unit contained several individual CpG sites (in 10 cases a CpG unit included 2 CpG sites, and in 2 cases 3 sites), the average methylation values were reported. Call rate regarding the 65 analyzable CpG units was 98.3%. Amplicon sizes, locations, and the number of CpG units covered are indicated in Table 1.

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Table 1

Details of amplicons included in this study

Amplicon ID	Length (bp)	No. of CpGs ^a	Location ^b		
			Genomic coordinates	Relative to TSS	
COMT_17	254	8	26:29366089 -29366342	-335 to -82	
COMT_26	194	6	26:29366881 -29367074	-1068 to -874	
HTR1A_14	204	4	2:50007473 -50007676	-1127 to -924	
MAOA_17	199	4	X:37676796 -37676994	-696 to -498	
MAOA_22	205	13	X:37677110 -37677314	-382 to -178	
TSS transcrip	tion start site				
^a Only number technology an		nalyzable by	MassARRAY [®] EpiT	TYPER [®]	
^b According to	o Ensembl an	notation Ca	nFam3.1 (GCA_0000	002285.2)	

Amplicon ID	Length (bp)	N C	Location ^b			
		No. of CpGs ^a	Genomic coordinates	Relative to TSS		
OXTR_17	246	6	20:9358073 -9358318	-124 to +122		
OXTR_34	148	4	20:9357391 -9357538	-805 to -658		
SLC6A4_8	164	4	9:44261258 -44261421	-1706 to -1543		
TPH1_2	470	4	21:40573179 -40573648	-2026 to -1555		
TPH1_7	359	5	21:40571913 -40572247	-624 to -266		
WFS1_15	289	15	13:38468824 -38469111	-1992 to -1703		
WFS1_32	198	7	13:38469484 -38469681	-2562 to -2365		
TSS transcrip	otion start site	e	I	I		

"Only numbers of CpGs analyzable by MassARRAY" Epil YPI technology are indicated

^bAccording to Ensembl annotation CanFam3.1 (GCA_000002285.2)

A CpG site (OXTR_17_CpG_3) with wide methylation range (18–52%) showing also significant inter-population methylation differences was selected for validation by pyrosequencing. Methylation levels measured by MassARRAY[®] significantly correlated with the results obtained from pyrosequencing (r = 0.95, p < 0.0001).

Characteristics and descriptive statistics of CpG units analyzed

Methylation levels and ranges largely varied with CpG unit (Fig. 1). 24.6% (16 of 65) of all CpG units analyzed showed minimal, i.e., \leq 5%, variation in methylation, while 44.6% (29 of 65) proved to be highly (\geq 20%) variable (Fig. 2a). Minimum observed variation in methylation levels was 1%, and the maximum was 68%. CpG units with \leq 5% variation in methylation fell in extreme methylation ranges (either >90 or <5%) with only a single exception, which actually consisted of 3 Globilities (WFS1_32_CpG_6.7.8; methylation range 76 -80%).

Fig. 1

Heat map showing the methylation ratios for each combination of samples and CpG units. CpG units are arranged along the x-axis and along samples are arranged the *y*-axis. COMT catechol-omethyltransferase, HTR1A 5-hydroxytryptamine receptor 1A, MAOA monoamine oxidase A, OXTR oxytocin receptor, SLC6A4 solute carrier family 6 member 4, TPH1 tryptophan hydroxylase 1, WFS1 wolframin endoplasmic reticulum transmembrane glycoprotein. G Golden Retriever, H Siberian Husky, B Border Collie, W wolf. Color code indicates percent methylation, ranging from 0.0% (light green) through 0.5% (black) to 1.0% (light red). Gray color indicates missing data values

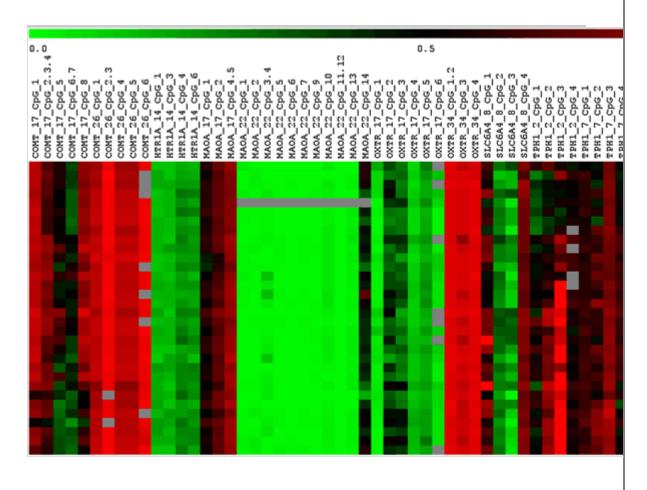
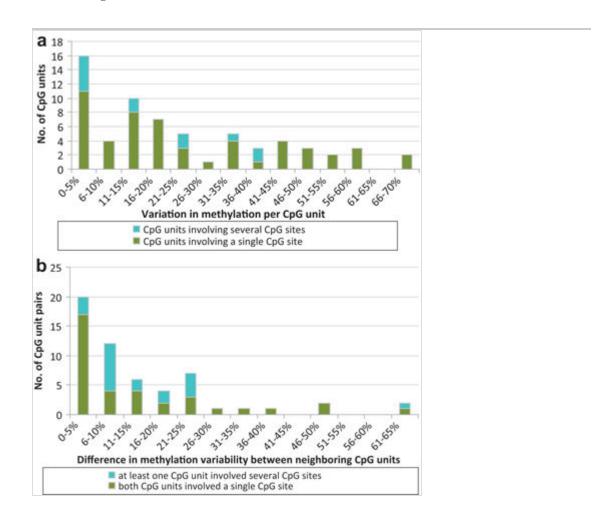


Fig. 2

Characteristics of the CpG units analyzed. CpG units comprising a single or several (2 or 3) individual CpG sites are indicated separately. **a**

http://eproofing.springer.com/journals/printpage.php?token=9TJDw81DY5-Mtt0ArK... 2017.03.20.

Distribution of the number of CpG units of a given variance range. **b** Difference between methylation variability* of neighboring CpG units. *Difference between methylation variability was measured by the formula: $|(H_A - L_A) - (H_B - L_B)| (H_A \text{ being the highest methylation}$ value detected for CpG unit A, L_A the lowest methylation value detected for CpG unit A, H_B the highest methylation value detected for CpG unit B, and L_B the lowest methylation value detected for CpG unit B)



Neighboring CpG units generally showed overlapping or nearly overlapping methylation ranges, with a \leq 5% difference between the highest detected methylation value for one CpG site and the lowest detected methylation value for its neighbor. Considering CpG units comprising a single CpG site only, those not fitting in these categories (e.g., those with non-overlapping methylation ranges where the difference between the highest detected methylation value for one CpG site and the lowest detected methylation value for one CpG site and the lowest detected methylation value for its neighbor was >5%) were observed in the case of only two out of 34 (5.9%) CpG neighbor pairs altogether. Difference in methylation variability was also generally low between neighboring CpG units, or, in other words, CpG units with a relatively wide methylation range were likely to have at least one neighboring CpG unit with wide methylation range as well, and vice versa (Fig. 2b). Using the formula $[(H_A - L_A) - (H_B - L_B)]]$ for expressing the difference between methylation variability (where H_A and L_A stand for the highest and lowest percentage methylation values observed for CpG unit A and accordingly, H_B and L_B stand for the highest and lowest percentage methylation values observed for the neighboring CpG unit B), a $\leq 5\%$ difference was observed in the case of 20 of all 56 (35.7%) CpG unit neighbor pairs, and altogether 32 (57.1%) CpG unit neighbor pairs showed a $\leq 10\%$ difference. Considering CpG unit pairs where both CpG units comprised a single CpG site only, this phenomenon was even more pronounced: 20 of all 36 (55.5%) such CpG site pairs showed a $\leq 5\%$ difference in methylation variability.

Differences in average amplicon methylation between populations

Average methylation values per amplicon were compared between the gray wolf and three purebred dog breeds: Golden Retriever, Siberian Husky, and Border Collie (Supplementary Table 2). Statistically significant (p < 0.05) inter-population differences were obtained in five of the 12 amplicons analyzed. Four of these, located in *COMT*, *TPH1*, and *WFS1* promoter regions, reached the level of significance of p < 0.0001. Wolf characteristic methylation values always score in the extreme range (either lowest or highest) (Supplementary Fig. 1).

CpG units with significant inter-population methylation differences

In order to gain a more detailed insight into population-specific methylation patterns, methylation statuses of each single CpG unit were also compared between the four populations. Tukey's test following one-way ANOVA showed significant differences between at least two canine populations in 52.3% (34 of altogether 65) of CpG units analyzed (Table 2). Wolves presented with the highest number of significant pair-wise differences (58 vs. 50, 45, and 37 in Golden Retrievers, Siberian Huskies, and Border Collies, respectively). Similarly, the highest number of extremely significant (p < 0.0001) pairwise CpG unit methylation differences was also obtained in the case of wolves (41 vs. 25, 24, and 20 in Golden Retrievers, Siberian Huskies, and Border Collies, respectively).

Table 2

CpG units with significant pair-wise inter-population methylation differences acc hoc test following one-way ANOVA

Region			Inter-group difference significant by Tukey				
Amplicon	CpG	ANOVA p	Golden Retriever vs. Siberian Husky	Golden Retriever vs. Border Collie	Golden Retriever vs. Wolf	Siberian Husky vs. Border Collie	
COMT_17	1	< 0.0001	No	No	Yes***	No	
	2-3 -4	0.0003	Yes**	No	No	No	
	5	< 0.0001	No	No	Yes***	No	
	6–7	< 0.0001	Yes***	Yes*	No	No	
	8	< 0.0001	No	No	Yes***	No	
MAOA_17	4–5	0.0274	No	Yes*	No	No	
MAOA_22	1	0.0204	No	Yes*	No	Yes*	
	2	0.0050	No	No	No	No	
	5	0.0163	Yes*	No	No	No	
	7	0.0163	Yes*	No	No	No	
	9	0.0158	No	No	No	No	
	10	0.0372	No	No	No	No	
	11 -12	< 0.0001	Yes***	No	No	Yes***	
OXTR_17	3	0.0010	No	No	Yes**	No	

0.001): three asterisks)

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Region			Inter-group difference significant by Tukey				
Amplicon	CpG	ANOVA p	Golden Retriever vs. Siberian Husky	Golden Retriever vs. Border Collie	Golden Retriever vs. Wolf	Siberian Husky vs. Border Collie	
	4	< 0.0001	No	Yes*	Yes***	No	
SLC6A4_8	4	0.0156	No	No	No	Yes*	
	1	0.0006	Yes*	Yes***	Yes*	No	
TDUI 0	2	< 0.0001	No	Yes***	Yes***	Yes***	
TPH1_2	3	0.0014	Yes***	Yes***	Yes***	No	
	4	0.0002	Yes***	No	No	Yes***	
TPH1_7	1	0.0053	No	Yes*	Yes*	No	
	2	0.0016	Yes*	No	Yes**	No	
	3	< 0.0001	No	No	Yes***	Yes*	
	4	< 0.0001	Yes***	Yes***	No	No	
	5	0.0038	No	No	Yes**	No	
	10	< 0.0001	Yes**	Yes**	Yes***	No	
WFS1_15	11	< 0.0001	Yes**	Yes***	Yes***	Yes***	
	12 -13	< 0.0001	No	No	Yes***	No	
	14	< 0.0001	Yes**	Yes*	Yes***	No	
	15 -16	< 0.0001	No	No	Yes***	No	
	17	< 0.0001	Yes*	Yes*	Yes***	No	
	18	< 0.0001	Yes**	No	Yes***	No	
	19 -20	< 0.0001	No	Yes**	Yes***	No	
WFS1_32	2	0.0319	No	Yes*	No	No	

Each row represents a CpG unit (consisting of 1–3 individual CpG sites). Level pair-wise difference in methylation is indicated by the number of asterisks (sigr 0.05): single asterisk; very significant ($0.001 \le p < 0.01$): two asterisks; highly ≤ 0.001): three asterisks)

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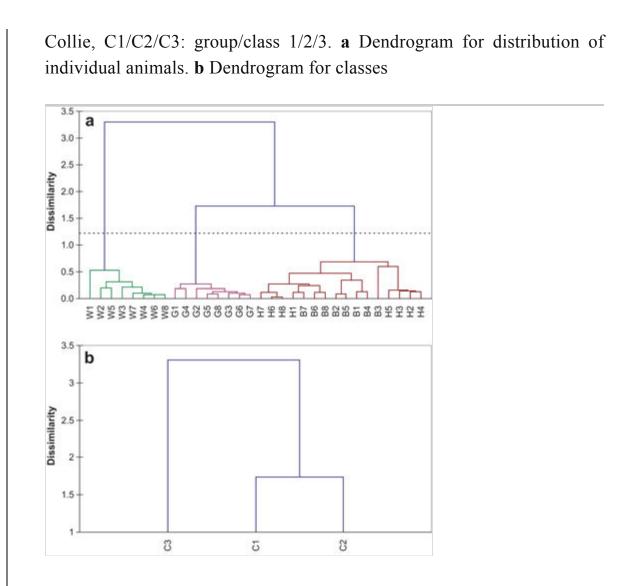
The vast majority (82.3%; 28 of 34) of CpG units showing significant (p < 0.05) pair-wise methylation differences between at least two populations emerged in CpG units with high, \geq 20% variation in methylation levels. On the reverse, significant inter-population differences were found in 72.4% (21 of 29) of all CpG units with \geq 20% variation in methylation. Considering CpG units with highly significant (p < 0.001) pair-wise methylation differences only, 90.9% (20 of altogether 22) of those showed \geq 20% variation in methylation, and highly significant inter-population differences were observed in 65.5% (19 of 29) of all CpG units with \geq 20% variation in methylation. Lowest degree of variation in methylation yielding significant pair-wise interpopulation differences was 11% (in OXTR_17_CpG_4), where the level of significance reached p < 0.001.

Cluster analysis

Hierarchical cluster analysis based on Ward's minimum variance between individual animals' methylation profiles (on the CpG unit level) indicated three distinct groups (Fig. 3). Central objects to groups/classes 1, 2, and 3 were Golden Retriever G3, Siberian Husky H6, and wolf W8, respectively. Distance between central objects of the classes was as follows: 0.736 for groups 1 and 2, 1.045 for groups 1 and 3, and 0.797 for groups 2 and 3, while the distance between the class centroids was as follows: 0.570 for groups 1 and 2, 0.984 for groups 1 and 3, and 0.671 for groups 2 and 3. Within-class variance regarding groups 1, 2, and 3 was 0.147, 0.242, and 0.214, respectively. Average distance to centroid by class was 0.351 (range 0.254–0.475), 0.463 (range 0.324–0.790), and 0.413 (range 0.281–0.681) for groups 1, 2, and 3, respectively. Variance decomposition for the optimal classification values was 58.78% for within-class variation and 41.22% for the between-class differences.

Fig. 3

Hierarchical cluster analysis based on CpG unit level methylation profiles. Euclidean distances of the three classes are shown together with pair-wise dissimilarities in methylation patterns regarding all animals investigated. W wolf, G Golden Retriever, H Siberian Husky, B Border



Segregation of wolf from domestic dog was unambiguous, as all wolves investigated belonged to a single cluster (group 3) containing no dogs at all and also showing a markedly higher degree of dissimilarity from both of the other two clusters than those from each other. Regarding dogs only, one of the three breeds (Golden Retriever) also formed a separate group (group 1); however, representatives of the two remaining breeds (Siberian Husky and Border Collie) were assigned to the same cluster (group 2).

Discussion

Here we provide supporting evidence for an altered epigenetic state of genes related to behavioral and cognitive functions between dog and wolf and, to a lesser extent, also between dog breeds. The findings are in harmony with genome-wide DNA methylation data on dog contra wolf (Janowitz Koch et al. 2016) as well as on natural darter

populations exhibiting different stages of evolutionary divergence (Smith et al. 2016), wild and domesticated chicken strains, and also human versus ape (Mendizabal et al. 2016). Yet, to our best knowledge, this is the first time when DNA methylation status has been demonstrated to provide a suitable basis for population assignment of individuals even with a relatively few and moderately variable markers, at least when promoter regions of a probable selection target gene group are interrogated.

Naturally, comprehensive interpretation of the findings would require further, extensive investigations. Perhaps most importantly, functional relevance of CpG sites showing population-specific differential methylation should be elucidated. Theoretically, promoter methylation levels of behavior-related genes could exert considerable influence on temperament and cognition, and thus the observed DNA methylation differences could contribute to the characteristic behavioral differences between the study populations. Methylation levels of promoter regions have often been shown to strongly correlate with gene expression levels (Portela and Esteller 2010). Also, characteristic, and apparently hereditary, gene expressional differences have been described in the brain of wild versus domesticated or tame strains of several species, including dog and wolf (Kukekova et al. 2011; Albert et al. 2012; Li et al. 2013), just as in aggressive contra non-aggressive dogs (Vage et al. 2010). Besides, DNA methylation together with other types of epigenetic modifications is a confirmed regulator of emotional behavior, behavioral memory, and synapse plasticity (Miller and Sweatt 2007; Miller et al. 2008, 2010; LaPlant et al. 2010; Yu et al. 2011).

However, presently it is uncertain whether methylation states of the here analyzed regions, especially as measured in buccal epithelial tissue, indeed correspond to gene expression and functioning. Epigenetic regulation is tissue specific, and hence behavior can only be influenced by DNA methylation patterns of the brain. Yet, some data point that methylation states, both global and CpG-wise, of a peripheral surrogate tissue often reflect effectively on that of the brain (Gregory et al. 2009; Thompson et al. 2013; Walton et al. 2016). Surprisingly, this can apply even when significant inter-tissue differences in RNA expression are present (Horvath et al. 2012). Human-related studies found dynamic changes in promoter methylation of stress-related genes, including the here analyzed *OXTR*, in white blood cells upon acute psychosocial stress (Unternaehrer et al. 2012), regardless of the fact that methylation of the human *OXTR* promoter had earlier been found to regulate gene expression in a tissue-specific manner (Kusui et al. 2001) and that *OXTR* expression in blood is negligible in comparison to brain according to the AceView annotation

(http://www.ncbi.nlm.nih.gov/IEB/Research/Acembly) (Thierry-Mieg and Thierry-Mieg 2006). Methylation states of *OXTR* promoter as measured in peripheral blood cells have been shown to associate with childhood maternal care (Unternaehrer et al. 2015), and *OXTR* promoter methylation levels in saliva correlated with anxiety/ depression (Chagnon et al. 2015). *OXTR* promoter methylation in blood has been found to closely correspond to neural response to ambiguous social stimuli in several brain regions according to functional magnetic resonance image (fMRI) scanning (Jack et al. 2012) as well. Although the other genes involved in this study are yet underinvestigated with regard to DNA methylation, especially in canines, studies on human *OXTR* indicate that data gained from surrogate tissues might be more valuable also from the functional aspect, than it could be assumed based on purely theoretical considerations.

Of the most commonly used peripheral tissues, buccal epithelial tissue seems to be an even better option than blood (Lowe et al. 2015; Smith et al. 2015) at least partly because of the unique methylation profile of the latter (Varley et al. 2013; Lokk et al. 2014). The fact that buccal epithelial cells are of the same embryologic ectodermic origin as neurons (Solnica-Krezel and Sepich 2012) provides a plausible biological ground as to why buccal methylation patterns could reliably reflect on that of the nervous system, especially if these patterns are already present at an early stage of embryonic development, just as it can be expected with inherited epigenetic marks. Presumably, in the course of tissue differentiation, cells and tissues derived from the same germ layer develop their own epigenetic patterns at varying rates and extents, so that in the adult organism high levels of inter-tissue DNA methylation correlation might be found where divergence rates were either low or where basically similar changes occurred. Indeed, it has been shown that DNA methylation levels can highly vary between

different brain areas (Harony-Nicolas et al. 2014; Hannon et al. 2015) and even between neuronal subtypes (Kozlenkov et al. 2016). Importantly, however, DNA methylation patterns were shown to be generally more homogeneous between different brain regions of the same individual than they are for the same brain region from different individuals (Illingworth et al. 2015). It is thus reasonable to assume a similar scenario for another ectoderm-derived tissue, justifying the usefulness of buccal epithelia for exploring population-specific methylation patterns. Nevertheless, it must be emphasized that only an exceedingly comprehensive, cell type-level inter-tissue DNA methylation analysis could provide a full picture, ideally with the inclusion of possible brain regions at all developmental stages and on a suitably large sample size to allow for statistical correction of multiple comparisons. Still, it might as well be that the nature of the observed relationship between buccal methylation states and population assignment lies somewhere else than brain expression levels. However, it seems unlikely that any epigenetic mark could show unambiguously population-specific patterns if it has no genuine biological relevance on gene expression levels.

Exploring the regulatory potential of the regions showing significant inter-population methylation differences would also be crucial. Unfortunately, experimental and observational data on the role of (putative) canine gene regulatory regions at present are extremely scarce. Of the here investigated genes, such information is only available with respect to MAOA (Eo et al. 2016). In this recent work, brain DNA methylation levels were investigated in three MAOA promoter segments together with mRNA expression levels in different breed dogs. Notably, one of the three promoter segments ("Region 1") overlapped precisely with the here analyzed amplicon MAOA 22. Similarly to our findings, marked differences were found between DNA methylation levels of the three breeds investigated, and there was a strong negative correlation between DNA methylation and gene expression levels. Indirect evidence also strongly supports the gene expression regulatory effect of MAOA 22 as well as of OXTR 17, since human orthologous fragments are indicated to be associated with the binding of several transcription factors and active chromatin marks including histone acetylation and DNAse I hypersensitivity clusters (the

latterivals with sypesal by the ENCODE

Consortium database (https://genome.ucsc.edu/) (Consortium 2012). Besides, human orthologous fragment of OXTR 34 was found to be associated with CTCF binding and DNAse I hypersensitivity clusters, and the direct neighborhood of the human orthologous fragment of HTR1A 14 was associated with transcription factor binding and histone acetylation. However, these data can by no means fully compensate for the lack of data on canines, and not even such an indirect evidence is available on most regions investigated in this study (namely COMT 17, COMT 26, MAOA 17, SLC6A4 8, TPH1 2, TPH1 7, WFS1 15, and WFS1 32) due to the complete lack of inter-species sequence homology at the investigated non-coding regions. Yet, it must be mentioned that online transcription factor binding predictor tools JASPAR (Sandelin et al. 2004; Mathelier et al. 2016) and PROMO (Messeguer et al. 2002; Farre et al. 2003) identified several putative transcription factor binding sites in all amplicons analyzed, irrespective of the level of phylogenetic conservation of their sequences. This fits in perfectly with the observation that the methylation status of promoter and nearpromoter CpG island shores, like the here investigated regions, often influences transcription efficiency of the proximal genes. In silico analyses thus support that the regions investigated might bear regulatory potential, even though this cannot be unambiguously stated in the lack of confirmatory experimental data.

Given the well-known influence of environmental factors on DNA methylation levels (Aguilera et al. 2010; Parle-McDermott and Ozaki 2011; Szyf 2011), the role of heredity versus environment in the observed population segregation is also to be elucidated. At present, it can only be stated that all dogs involved in the study, independent of breed, were family dogs living in different households, and thus environmental differences as causative factors in the segregation of Golden Retriever from the other two dog breeds seem unlikely. With regard to wolves, the situation is slightly complex. Clearly, they were not family pets, but their keeping conditions had many in common with those of dogs and markedly differed from those of free-ranging wild wolves. They were hand-raised, had regular contact with humans also in adulthood, and underwent clicker training similar to dogs. Besides, they constituted a heterogeneous population with regard to social effects, as they were members of separate packs and occupied different ranks in the hierarchy. Yet, they formed a uniform cluster in terms of DNA methylation, implying that the role of environment might be minor compared to the role of heredity in their segregation from dog.

Of similar concern is the inclusion of North American timber wolves instead of a Eurasian wolf population. Although the exact location of the domestication center is still disputed (vonHoldt et al. 2010), it is agreed upon that the domestic dog evolved in the Eurasian continent, most possibly in Southeast Asia (Savolainen et al. 2002; Ding et al. 2012; Thalmann et al. 2013). This naturally raises the question whether the observed dog-wolf DNA methylation differences might rather reflect wolf phylogeny than dog domestication and artificial selection. Yet, North American timber wolves might not represent a particularly more remote population from dog than any modern Eurasian wolves. Evidence indicates that the domestic dog derived from a founder population with a fairly larger genetic diversity than observed among any modern wolf populations (Freedman et al. 2014). Besides, the most careful estimations put dog domestication at about 11,000–14,000 years ago (Axelsson et al. 2013; Freedman et al. 2014), with some at about already 33,000 years ago (Germonpre et al. 2009; Wang et al. 2016). North American timber wolves are believed to represent the last of three separate invasions into North America at about 11,000–12,000 years ago from Eastern Asia through the Bering land bridge (Vila et al. 1999), implying that the ancestors of North American timber wolves could have been similarly closely related to those of dogs as ancestors of modern Eurasian wolves themselves.

Another important issue would be to explore whether the findings also apply for non-brain-related genes, non-promoter regions, and epigenetic modifications other than DNA methylation (i.e., histone modifications and non-coding RNA species). Evidence suggests that the answer to at least the first two questions is yes, although the extent of interpopulation DNA methylation differences might considerably vary with gene function and region. A recent study investigating DNA methylation in dog and wolf blood samples using reduced representation bisulfite sequencing (Janowitz Koch et al. 2016) found that domestication-associated differentially methylated CpGs were highly enriched for repeat regions and included several functionally relevant gene ontology groups, including, but not restricted to, genes involved in neurobiological processes. Notably, this study also found that dogs and wolves formed different clusters based on methylation data, even though population subdivision varied with the analysis method. Yet, unpublished data of our group on DNA methylation in promoter regions of 20 genes in mixed-breed dogs and wolves living under the same conditions also indicate the existence of dog- and wolfspecific methylation patterns, but at the same time results also hint that this phenomenon is dependent on the gene categories investigated. It must also be noted that although all genes analyzed in the present study were previously shown to affect behavior, not all of them are classical behavior-related genes. Wolframin gene (WFS1) encodes an endoplasmic reticulum transmembrane protein and is expressed bodywide (De Falco et al. 2012), with its highest expression levels being reported in non-brain tissues. Besides, even genes typically associated with neurosignaling including COMT encoding catechol-O-methyltransferase or TPH1 encoding tryptophan hydroxylase are expressed in several non-brain tissues at levels similar to those observed in the brain, making the picture even more complex. Importantly, comparative brain methylome analysis on chicken domestication (Natt et al. 2012) indicated that inter-population differentially methylated regions are enriched in genes related to cell signaling, thereby influencing stress tolerance, cognitive functions, and reproduction-features that are important from the domestication perspective. It was also shown that epigenetically affected genes were over-represented in selective sweep regions. Yet, a muscle methylome analysis on chicken strains found a large degree of conservation instead of divergence in promoter methylation states (Li et al. 2015), implying that population-specific epigenetic shifts do not occur randomly. In concert with these findings are the results gained from larger evolutionary scale methylome comparison studies between human and ape, with regard to both brain and non-brain tissues (Molaro et al. 2011; Pai et al. 2011; Mendizabal et al. 2016). These studies also highlighted that inter-species differentially methylated regions can be identified at least as frequently in extra-promoter regions (i.e., in gene bodies and intergenic regions) as within promoters (Mendizabal et al.

2016), similar to tissue-specific differentially methylated regions (Lokk et al. 2014).

The observation that also CpG sites with modest degrees of methylation variation can show population-specific methylation differences would also require further validation and elucidation. Similarly, extensive investigations would be needed for estimating the occurrence and determining the significance of differentially methylated CpGs located in regions that otherwise show low levels of methylation variability and lack population-specific patterns. General observations highlight that the methylation levels of neighboring CpG sites tend to fall into overlapping ranges, and genomic regions of marginally high and low levels of methylation are typically separated by an intermittent zone, where the methylation levels of neighboring CpG sites gradually change (Mikeska et al. 2007; Couldrey and Lee 2010). Our results also demonstrated that the neighboring CpG sites often relatively show differences in the extent of methylation variability, rendering it challenging to predict on the characteristics of a single CpG site in the absence of referring experimental and observational data.

In summary, many open questions remain with regard to full interpretation of the results. Yet, despite all the limitations of this study, the perspectives offered by the findings are intriguing. Apparently, epigenetic divergence of populations can be specific to the level making population assignment of individuals as well as phylogeny reconstruction possible purely based on epigenetic marks, at least when regulatory regions of putative selection target genes are interrogated. Apart from the direct practical aspects of these perspectives, the findings also add to the growing body of evidence that epigenetic changes accompany, if not downright drive, population segregation processes.

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Compliance with ethical standards

Conflict of interest All the authors declare no conflict of interest.

Ethical approval All applicable international, national, and/or institutional guidelines for the care and use of animals were followed. This article does not contain any studies with human participants performed by any of the authors.

Electronic supplementary material

Below is the link to the electronic supplementary material.

Supplementary Fig. 1: Percent methylation of amplicons with p = 0.0001 inter-population differences in the four populations investigated. Methylation values for each individual animal are indicated. Horizontal lines represent mean values \pm SD (standard deviation) (TIF 42932 KB)

Supplementary Table 1 (DOC 47 KB)

Supplementary Table 2 (DOC 103 KB)

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