The extinct Sicilian wolf shows a complex history of isolation and admixture with ancient dogs

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Schematic representation of the Sicilian wolf evolution

- Sicily-Italy landbridge (Antoniol et al. 2016)
- LCA of the Sicilian and Italian wolf arrives in Sicily
- Sicilian wolves and ancient European dogs inferred admixture time
- Sicilian wolf extinction
- Modern wolf populations coalescent time range (Fan et al 2016)
- European dog lineage replacement (Bergstrom et al. 2020)
- Sicilian wolves and modern European dogs inferred admixture time

European wolves
Italian wolf
Sicilian wolf
Modern dog breeds (Europe)
Eneolithic Bronze Age dogs (Europe)

Highlights
- The Sicilian and Italian wolves derive from the same evolutionary lineage
- Sicilian wolf genomes carry Eneolithic dog ancestry
- Nearly half of the Sicilian wolf genome is in runs of homozygosity
wolves' evolutionary history.5,7,9 wolves and domestic dogs, as well as recurrent post-divergence gene flow events, are common features of disturbance of their habitat.5,6

Introductions, which suggests a decline in population connectivity and gene flow potentially due to anthropogenic

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The gray wolf (Canis lupus) is one of the most widespread and mobile carnivores in the Holarctic region.1 Genomic studies have grouped modern wolves into two distinct phylogenetic lineages: the Eurasian and North American wolf.1 Recent palaeogenomic studies demonstrated that Eurasian wolf populations were highly connected throughout the Late Pliocene but experienced a significant turnover after the Last Glacial Maximum (LGM).2–4 Today, gray wolf genetic diversity surveys show they form structured populations, which suggests a decline in population connectivity and gene flow potentially due to anthropogenic disturbance of their habitat.5,6

Albeit the deep evolutionary relationship between humans and wolves, their direct competition and human persecution have led wolf populations across their geographical distribution to population decline, geneti
c bottlenecks, and even local extinctions.7,8 Therefore, changes in population size, admixture between wolves and domestic dogs, as well as recurrent post-divergence gene flow events, are common features of wolves’ evolutionary history.5,7,9

The Italian peninsula was the theater of peculiar population dynamics during LGM. It was a glacial refugium profoundly shaped by sea-level changes. In addition to the recognized local wolf subspecies (C. l. italicus),10–13 Italy harbored one of the last Mediterranean wolf populations, confined only in Italy until the mid-1900s.14 The Sicilian wolf remained isolated on the island from the end of the Late Pleistocene (ca. 17kya),15,16 until its human-mediated extermination in the first half of the twentieth century (the 1930s–1960s).17–20 Today, there are only seven known Sicilian wolf specimens from the 19-20th centuries preserved in museums around Italy. Morphological measurements obtained from the specimens stored in museums and from historical records indicate that this insular population was morphologically distinct from wolves in continental Europe.14,21,22 The Sicilian wolf’s most peculiar features were its small size,
the fur tending to light yellow, and the absence of the dark stripes on the forearms, a distinctive character of the peninsular Iberian wolf subspecies. The minute size of the Sicilian wolf places it among the smallest subspecies of gray wolf, together with the Arab wolf (C. l. arabs), and the extinct Japanese wolf (C. l. hodophila). Based on these morphometric analyses, the Sicilian wolf has been assigned to the wolf subspecies Canis lupus cristaldi. However, despite the modern Italian wolf population being morphologically and genetically well characterized, few genomes are available from ancient and insular Italian samples and the genomic identity of the Sicilian wolf remains debated.

To investigate the Sicilian wolf’s genomic ancestry and the potential consequences of the long-term isolation in its genome, we sampled the seven existing museum specimens and were able to recover four nuclear genomes and five mitogenomes (Table S1).

RESULTS AND DISCUSSION

Reconstructing the genomes of the Sicilian wolves

We sampled the seven existing museum specimens of the Sicilian wolf, from which we recovered four nuclear genomes and five mitogenomes (Table S1). Poor DNA preservation hindered the recovery of genetic material from two of the samples. We achieved an average depth of coverage ranging between 3.8–11.6× for the nuclear genomes and 19.7–1239.2× for the mitochondrial genomes (Table S1). Mapped reads displayed an increase of C to T substitution rates at the 3’ end of the reads, characteristic of sequencing data from historical specimens (Figure S1). For comparative purposes, we sequenced 33 modern wolf genomes (3.6–41.9×) from across Europe and 3 Cirneco dell’Etna dogs (2–2.6×), an old Sicilian hunting breed. We combined our new data with publicly available genomes from dogs and wolves. The final dataset consisted of 57 modern Eurasian and 5 American wolves, 5 Sicilian wolves, 5 Pleistocene wolves, 45 modern dogs, 24 ancient dogs, 4 coyotes (Canis latrans), and 1 Andean fox (Lycalopex culpaeus) (Table S2). The coyote and Andean fox were used as outgroups.

Genetic affinities of the Sicilian wolf

We investigated the placement of the Sicilian wolves in the context of global dog and wolf genomic diversity by performing a multidimensional scaling (MDS) analysis including the Sicilian wolves, modern Eurasian wolves, modern North American wolves, and Pleistocene Siberian wolves (Figure 1A) in the resulting plot, the first dimension (Dim1 - 15.42%) separated dogs from wolves, while the second dimension (Dim2 - 3.62%) separated different groups of wolves, with Asian Highland and Italian wolves at each end of the distribution. In the MDS, the Sicilian wolves are placed within the European wolf diversity in proximity to the Italian but shifted toward the cluster of dogs (Figure 1A). An MDS analysis restricting to Eurasian wolves separated the Sicilian wolf from the rest of the wolves in dimension 1, while separating the Italian and Sicilian wolves from the Highland wolf in dimension 2 (Figure S2A). The clear separation of the Sicilian wolves from all other groups in the MDS is consistent with an island population with a history of long-term isolation.

To further explore the ancestry and potential admixture in the Sicilian wolf genomes, we used the model-based clustering approach implemented in ADMIXTURE v1.3 (Figures 1B, S2B and S2C). When assuming two ancestry components (Ks), the samples are mostly separated into dogs and wolves, with the Sicilian wolves being modeled as a mixture of both components. This dual admixture pattern was not exclusive to the Sicilian wolves. However, the Sicilian wolf displayed the largest proportion of the dog-related ancestry component, only after the Sierra Morena wolf, which derives one-third of its genome from dog ancestry. To further explore the ancestry and potential admixture in the Sicilian wolf genomes, we used the model-based clustering approach implemented in ADMIXTURE v1.3 (Figures 1B, S2B and S2C). When assuming two ancestry components (Ks), the samples are mostly separated into dogs and wolves, with the Sicilian wolves being modeled as a mixture of both components. This dual admixture pattern was not exclusive to the Sicilian wolves. However, the Sicilian wolf displayed the largest proportion of the dog-related ancestry component, only after the Sierra Morena wolf, which derives one-third of its genome from dog ancestry. To further explore the ancestry and potential admixture in the Sicilian wolf genomes, we used the model-based clustering approach implemented in ADMIXTURE v1.3 (Figures 1B, S2B and S2C). When assuming two ancestry components (Ks), the samples are mostly separated into dogs and wolves, with the Sicilian wolves being modeled as a mixture of both components. This dual admixture pattern was not exclusive to the Sicilian wolves. However, the Sicilian wolf displayed the largest proportion of the dog-related ancestry component, only after the Sierra Morena wolf, which derives one-third of its genome from dog ancestry. To further explore the ancestry and potential admixture in the Sicilian wolf genomes, we used the model-based clustering approach implemented in ADMIXTURE v1.3 (Figures 1B, S2B and S2C). When assuming two ancestry components (Ks), the samples are mostly separated into dogs and wolves, with the Sicilian wolves being modeled as a mixture of both components. This dual admixture pattern was not exclusive to the Sicilian wolves. However, the Sicilian wolf displayed the largest proportion of the dog-related ancestry component, only after the Sierra Morena wolf, which derives one-third of its genome from dog ancestry.

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The dual wolf-dog ancestry of the Sicilian wolf

We further explored whether the patterns observed in the MDS and ADMIXTURE analyses were due to the Sicilian wolf being a highly drifted population or dog admixture. We used outgroup $f_3$-statistics, which measure the shared genetic drift between pairs of samples, to identify the closest populations to the Sicilian wolf. We found that the Sicilian wolf was most closely related to ancient and modern European dogs, followed by modern Italian wolves (Figure 2A). Among modern dogs, the Italian dogs, such as Spinone, Cane Corso, and Cirneco, had the highest genetic affinity to the Sicilian wolf (Figures 2A and S3A–S3D). Recent introgression of stray dog ancestry present on the island could explain the high affinity between the Sicilian wolf and modern Italian dogs. However, the high affinity between the Sicilian wolf and ancient Eneolithic and Bronze Age dogs (â€”5000–3000 BP; e.g., Croatian ALPO01, Italian AL2397, Irish Dog-1PU, and Newgrange) suggests a more ancient origin for this link. We observed a similar pattern in all four Sicilian wolf genomes, with the Sicilian wolf 1 being particularly close to ancient European dogs (Figures 2A and S3A–S3D). In contrast, outgroup $f_3$-statistics for the Italian and Iberian wolves, which are the closest to the Sicilian wolf (Figure 1) and share a similar history of recent isolation and bottlenecks, did not show the same pattern (Figures S3E and S3F).

We then used D-statistics to formally test for gene flow between the Sicilian wolves and modern and ancient dogs. We tested if the Sicilian wolf shared more alleles with dogs compared to its closest wolf (the Italian wolf) (Figures 2A and 2C). Our results showed that the Sicilian wolf shares significantly more alleles with dogs compared to its closest wolf (Z score >3.33). Furthermore, ancient European dogs yielded the largest D values (Figure 2B), confirming that ancient European dogs are the closest to the Sicilian wolf among dogs. A similar result was obtained for all four Sicilian wolf genomes (Figures S4A and S4B). However, when comparing the D-statistics obtained for different pairs of Sicilian wolves, we found each has a different degree of affinity to dogs suggesting they have varying dog ancestry proportions. Two scenarios could explain

Figure 1. Population structure of the Sicilian wolves retrieved from genomic data
(A) Multidimensional scaling plot showing the genomic diversity of the Sicilian wolf, ancient and modern wolves, and dogs. Dimensions one (Dim1 - 15.42%) and two (Dim2 - 3.62%) are represented on the x and y axes. Each point indicates a different sample, and colors indicate the broad geographic categories in the legend.
(B) ADMIXTURE plots assuming two (top) and five (bottom) ancestry components. Vertical bars correspond to different samples, and the colors represent the estimated ancestry components and proportions. Horizontal bars in different colors below the ADMIXTURE results indicate the main geographical and temporal groups.

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these observations: gene flow (ancient, recent, or both) from dogs into the Sicilian wolf, or the Sicilian wolf being closer to a now extinct (and unsampled) wolf lineage that contributed to ancient European dogs. While the second scenario is supported by mtDNA based phylogenetic studies,\textsuperscript{27,28} which suggest genetic continuity between Late Pleistocene wolves in Italy and dogs, our results support the first scenario. The \textit{D}-statistics showed significant gene flow between the Sicilian wolf and dogs, regardless of the dog used in the test (Figures 2Ba and S4A). If the gene flow direction were from the Sicilian wolf into specific dog breeds, we would expect to observe significant values only for those dog breeds carrying Sicilian wolf ancestry. Furthermore, these results were consistent when using the Portuguese wolf instead of the Italian wolf in the \textit{D}-statistic test (Figure S4E). Stray dogs have been common in Sicily, potentially since their introduction to the island, making it possible that the dog ancestry in the Sicilian wolf could derive from recent admixture. We used \textit{D}-statistics to determine whether an ancient or modern dog was a better source for the dog ancestry in the Sicilian wolf. We computed \textit{D} (Eneolithic Croatian dog, X; Sicilian wolf, Andean fox) to estimate the amount of shared derived alleles between the Sicilian wolf and the Eneolithic Croatian dog (ALPO01) compared to all other dogs in the dataset (X). Then, we computed the same test substituting the Eneolithic Croatian dog for the Cirneco dog and compared the results of both tests (Figure S4B). We found that in most of the tests, the \textit{D}-statistic obtained with the Eneolithic Croatian dog is larger than that obtained with the Cirneco dog, suggesting the former is a better source for the dog ancestry of the Sicilian wolf.

Figure 2. Ancestry composition of the Sicilian wolf
(A) Outgroup \textit{f}_3-statistics of the form \textit{f}_3(Andean fox; Sic1, X), showing the genetic affinity between the Sicilian wolf (Sic1) and other dogs and wolves (X). Names are indicated for the samples with the highest affinity to the Sicilian wolves (highest \textit{f}_3). Error bars correspond to 3.3 standard errors estimated through a block jackknife approach. Corresponding plots for the other Sicilian wolf genomes are shown in Figure S3.
(B) \textit{D}-statistic tests of the form \textit{D}(Sic1, Italian wolf, H3, Andean fox) (x axis) and \textit{D}(Sic7, Italian wolf, H3, Andean fox) (y axis), where H3 corresponds to different groups of dogs and wolves. Each point corresponds to a different test where H3 varies. Error bars correspond to 3.3 standard errors estimated through a block jackknife approach. All tests yielded significant results (Z score >3.3) suggesting gene flow between the Sicilian wolf and the corresponding dog in each test. A dotted line indicates the identity line, with points above the line showing tests where the Sicilian wolf had higher affinity to H3 compared to Sicilian wolf 1.
(C) TreeMix admixture graph incorporating 3 migration edges (arrows): from ancient southern European dogs into the Sicilian wolf, from a Mesolithic Swedish dog (C89) into the Alaskan Malamute, and from the Middle eastern dog into ancient southern European dogs. The corresponding TreeMix residuals are shown in Figure S5.
To place the Sicilian wolves in the broader phylogenetic context of modern and ancient wolves and dogs while allowing for potential admixture events, we used TreeMix. The obtained graph without migration edges was consistent with a nuclear phylogeny estimated using RAxML (Figure S2D), placing the Sicilian wolves as intermediate between the dog and wolf clades (Figure S5A). Once migrations were incorporated, the Sicilian wolves were modeled as a sister clade to the Italian wolves with ancestry contribution from the Croatian Eneolithic dog lineage into the base of the Sicilian wolf clade (Figures 2C and S5). These results were consistent with F-statistics-based admixture graphs that showed the Sicilian wolf can be modeled as a mixture of ancient dog (~31–49%) and Italian wolf (~51–69%) ancestries (Figures S4C and S4D).

To further validate our results and to measure the extent of the dog ancestry in the Sicilian wolves, we performed local ancestry inference using PCAdmix. We modeled each Sicilian wolf genome as a mixture of dog and wolf ancestry using nine dogs and ten wolves as sources. Our results showed that the Sicilian wolves carry between ~50-60% wolf and ~50-40% dog ancestry (Figure S6), in agreement with the admixture graph modeling (Figures S4C and S4D). Altogether, these results showed that the Sicilian wolf is a mixture of two ancestries: wolf (most similar to the modern Italian wolf) and dog (most similar to ancient European dogs).

In light of our results, we hypothesize that geographic isolation, a small population size, and the introduction of dogs to Sicily, provided a fertile ground for hybridization and retention of ancient dog ancestry in the Sicilian wolf. A previous study on Sierra Morena wolves (C. l. signatus) found that one-third of their genomes were of dog ancestry. The Sierra Morena and the Sicilian wolves have similar demographic histories, being a small, isolated, and declining population in a human-modified landscape. A similar pattern of admixture with modern dogs was documented in one of the last specimens of Japanese wolves (C. l. hodophilax), which became extinct at the same time as the Sicilian wolf. Thus, our findings further show that admixture with dogs might be favored in low-density isolated populations in anthropogenic contexts. The affinity between Sicilian wolves and ancient European dogs represents the most striking finding, as it shows that ancient dog ancestry prevailed until recently in the Sicilian wolf genome. Overall, our results showed that the dual ancestry of the Sicilian wolf genome resulted from a long-term multi-factor evolutionary process involving geographic isolation and admixture with dogs, potentially influenced by a drastic population reduction due to resource competition and direct conflict with humans.

The effects of island isolation on the Sicilian wolf
We explored the insular effect and the consequences of the population decline in the last few hundred years on the genomic diversity of the Sicilian wolf genome. We estimated the average nucleotide diversity of the Sicilian wolf and European wolves for comparative purposes. The Sicilian wolf displayed low-average nucleotide diversity compared to the Italian, Iberian, and Scandinavian wolves (Figure 3A). Similarly, heterozygosity estimates for the two highest coverage Sicilian wolf genomes were low (Figure 3B). A decrease in diversity is a characteristic of island populations. Therefore the Sicilian wolf genomic diversity is consistent with the expectations from a population that was shaped by a strong founder effect, long-term isolation, and small effective population size.

To get insights into the distribution of low-heterozygosity regions in the Sicilian wolf genome we estimated runs of homozygosity (ROHs) for the two highest coverage samples. ROHs abundance and their length can be informative about population sizes and inbreeding. Our results showed that 37–50% of the Sicilian wolf genome can be found in ROHs with average ROHs lengths of 9.5Mb and 6.6Mb for Sicilian wolf 1 (Sic1) and 2 (Sic2), respectively (Figures 3C–3E). While long ROHs (>10Mb) were still quite prevalent in the Sicilian wolves (15 and 30% from total ROHs), the number of short and medium ROHs was higher than those observed in recently inbred populations such as the Mexican wolf (C. l. baileyi) (Figures 3C and 3D). We found the Sic2 had more ROHs compared to the Sic1. Notably, our D-statistic tests showed the Sic1 had a higher proportion of dog ancestry compared to Sic2. Admixture between two distinct populations leads to an increase in heterozygosity, thus it is possible that the smaller number of ROHs in Sic1 is due to its higher proportion of dog ancestry. Compared to the Italian wolf, its genetically closest wolf, the Sicilian wolf had a larger number of ROHs (Figure 3C) but their size distribution was similar (Figure 3D). A larger proportion of shorter ROHs could be the result of long-term reduction in population size of the Sicilian wolf, combined with influx of ancestry from dogs, which would break the larger ROH blocks.

To measure the potential impact of the small population size in the fitness of the Sicilian wolf compared to mainland populations we estimated their homozygous transversion load. The Sicilian wolf had a relatively
higher mutation load compared to other wolves (particularly Sic2), higher than the mutation load in the inbred Mexican wolf (Figure 3E). Combined, these results indicate that the Sicilian wolf population was suffering from the effects of small population size already at the end of the 19th century. Finally, our results showing the Sicilian wolf carried ancient dog ancestry together with the short ROH segments, suggest the admixture happened in the past, rather than toward the time of extinction.

**Conclusion**

The genomes of some of the last surviving Sicilian wolves showed that their evolutionary history was profoundly shaped by admixture with dogs. Our results showed they derive from the same lineage as modern Italian wolves but carry ancient Eneolithic and Bronze Age dog ancestry. Such ancient dog ancestry was shown to be widespread in Europe until 3-4 kya. Our results suggest this ancient dog lineage survived in the Sicilian wolf until the 20th century. In this respect, the Sicilian wolf is unique among all other extant and extinct European wolf populations genetically characterized to date. The phylogenetic placement of the Sicilian wolf placed together with the Italian wolf (Figure 2C), coupled with the evidence for long-term isolation, supports that by 25-17 kya (the time of the colonization of Sicily), the Italian and Sicilian wolves’ ancestor had already split from other European wolf lineages. Previous genomic studies estimated that contemporary wolf populations share a common ancestor between 32 and 13 kya 1,29,45 Our results provide a lower limit for the split of wolves from their last common ancestor, suggesting that contemporary populations were already differentiated by the time of the colonization of Sicily (25 kya). Finally, these results

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**Figure 3. Island isolation impact on the Sicilian wolf genomes**

(A) Population level nucleotide diversity of Sicilian, Italian, Iberian, and Scandinavian wolves. The chromosomes and the nucleotide diversity values are shown on the x and y axis respectively. The red line represents the mean diversity value.

(B) Heterozygosity in sliding windows (excluding repetitive regions) for different wolf samples show reduced levels of heterozygosity in the Sicilian wolves Sic1 and Sic2.

(C) Individual runs of homozygosity represented as the number of ROH segments on the y axis and fraction of the genome contained in ROH (Mbp) on the x axis.

(D) Histogram showing the distribution of ROH segments of different size categories. The Sicilian wolves (Sic1 and Sic2) show a high number of short ROHs.

(E) Genetic load represented as the number of transversion sites that are homozygous is shown on the y axis compared to the proportion of the genome in ROH (FROH). The Sicilian wolves display a higher transversion load compared to other wolf populations.
clarified previous hypotheses based on mitochondrial haplogroups, suggesting the Sicilian and Italian wolves form a clade with the Pleistocene Siberian wolves. Instead, our f3-statistic and MDS analyses showed that Siberian Pleistocene wolves were not closely related to the Sicilian wolf (Figures 2A and S3A–S3D).

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SUPPLEMENTAL INFORMATION
Supplemental information can be found online at https://doi.org/10.1016/j.isci.2023.107307.

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AUTHOR CONTRIBUTIONS


DECLARATION OF INTERESTS

The authors declare no competing interests.

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along each chromosome in individuals with admixed ancestry from two or more populations. Hum. Biol. 84, 343–364.


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RESOURCE AVAILABILITY

Lead contact
Further information and requests of resources and reagents should be directed to and will be fulfilled by the lead contact, Marta Maria Ciucani (ciucani@palaeome.org).

Materials availability
This study did not generate new unique reagents.

Data and code availability
- Sequencing data generated in this study have been deposited at the European Nucleotide Archive (ENA) under accession number PRJEB57290 and are publicly available as of the date of publication.
- This paper does not report any original code.
- Any additional information required to reanalyse the data reported in this paper is available from the lead contact upon request.

METHOD DETAILS

Description of the historical Sicilian samples
The sample Sic1 (cat number C11875) was a petrous bone from a wolf skull preserved at the Museum of Natural History, Section of Zoology ‘La Specola’, University of Florence, Florence, Italy. The sample still has the original tag from the 19th century, and it belongs to a wolf killed in 1883 at Vicari (Palermo, PA). The sample Sic2 (cat number AN/855) is represented by a petrous bone from a wolf skull belonging to a juvenile individual from 1879 preserved at the Museum of Zoology ‘Pietro Doderlein’ of the University of Palermo, Palermo, Italy. The sample Sic3 (cat number M/18) is a mounted specimen of an adult male. The individual Sic2 was killed in Sicily, presumably around 1870-1880, as documented by an old picture in the museum records. The specimen is preserved in the Museum of Zoology ‘Pietro Doderlein’ of the University of Palermo, Palermo, Italy. The individual Sic4 (cat number M/18) is a mounted specimen of an adult male. The individual Sic2 was killed in Sicily, presumably around 1870-1880, as documented by an old picture in the museum records. The specimen is preserved in the Museum of Zoology ‘Pietro Doderlein’ of the University of Palermo, Palermo, Italy. The last sample, Sic7 (cat number NA), is a hide dated around 1880-1920 and is preserved at the Museum of Termini Imerese (Palermo, PA), Italy.

Data generation for the historical Sicilian samples
The Sicilian wolf samples were processed under strict clean laboratory conditions at the Globe Institute, University of Copenhagen. DNA extractions were performed following the silica-based protocol described in Dabney et al. (2013) for the bone samples and Campos & Gilbert (2009) for keratin samples (hides and claws). Both bone and keratin samples were digested using 1 mL of digestion buffer. The extracts were

Continued
purified using modified PB buffer (Qiagen), washed twice with PE buffer (Qiagen) and eluted twice in 20 μL of buffer EB - with 10 minutes of incubation time at 37°C. The concentration of each extract was checked on a Qubit Fluorometer (ThermoFisher Scientific) in ng/μL and on the Tapestation (Agilent Technologies) for concentration and fragment size.

BGI libraries for the samples Sic1 and Sic4 were constructed following Carøe et al. (2018)88 and Mak et al. (2017)64,69, while BEST Illumina libraries for Sic2, Sic3 and Sic7 were built following Carøe et al. (2018) using Illumina adapters.88 Libraries were prepared using up to 32 μL of DNA in a final reaction volume of 80 μL.

The appropriate number of cycles for the amplification was determined using Mx3005 qPCR (Agilent Technologies) in which 1 μL of SYBRgreen fluorescent dye (Invitrogen, Carlsbad, CA, USA) was loaded in 20 μL indexing reaction volume also using 1 μL of template, 0.2 mM dNTPs (Invitrogen), 0.04 U/μL AmpliTaq Gold DNA polymerase (Applied Biosystems, Foster City, CA, USA), 2.5 mM MgCl2 (Applied Biosystems), 1X GeneAmp® 10X PCR Buffer II (Applied Biosystems), 0.2 μM forward and reverse primers mixture,64 and 16.68 μL AccuGene molecular biology water (Lonza, Basel, CH). qPCR cycling conditions were 95°C for 10 minutes, followed by 40 cycles of 95°C for 30 seconds, 60°C for 30 seconds, and 72°C for 45 seconds.

The libraries were amplified using PfTurbo Cx HotStart DNA Polymerase (Agilent Technologies) and Phusion® High-Fidelity PCR Master Mix with HF buffer (New England Biolabs Inc). The amplification was performed in 50 μL PCR reactions that contained:

- 14 μL of purified library, 0.1 μM of each forward (BGI 2.0) and custom made reverse BGI primers, 2X Phusion® High-Fidelity PCR Master Mix with HF buffer and 8.6 μL AccuGene molecular biology water.
- 10 μL of purified library, forward and reverse primers, 2.5U/L PfTurbo Cx HotStart DNA Polymerase, 0.4 BSA, 1 μL of dNTPs (25 μM), 1 μL of Buffer 10X and 30.6 μL AccuGene molecular biology water.

PCR cycling conditions for libraries amplified using Phusion® High-Fidelity PCR Master Mix with HF buffer were: initial denaturation at 98°C for 45 seconds followed by 18 to 20 cycles of 98°C for 20 seconds, 60°C for 30 seconds, and 72°C for 20 seconds, and a final elongation step at 72°C for 5 minutes. Amplified libraries were then purified using 1.5x ratio of SPRI beads to remove adapter dimers and fragments smaller than 90-100bp and eluted in 50 μL of EB (Qiagen) buffer after 10 minutes of incubation at 37°C. The samples Sic1 and Sic4 were sequenced using BGiseq platform, while the samples Sic2, Sic3 and Sic7 were pooled together and sequenced on Novaseq6000 Illumina platform, S2 flow-cell PE50.

Data generation for modern samples

Cirneco dell’Etna buccal swab samples - DNA sample collection, storage and extraction

SK-15 Isohelix buccal swabs for non-invasive DNA collection were used to sample three Cirneco dell’Etna individuals (MW1085_CIR1, MW1095_CIR11, MW1098_CIR14). The dogs were sampled by their owners at least 30 minutes after eating, and to avoid contamination, gloves and face masks were used. The swabs were inserted into the dogs’ mouths, rubbed on their cheek for ca. 1 minute and placed back into the collector tube with the SGC-50 Isohelix Dri-capsule to preserve and stabilise the DNA at room temperature during shipping. Once received in the Modern DNA labs of the Globe Institute (University of Copenhagen), the samples were stored at -20°C.

Buccal swab samples were placed in 2 mL Eppendorf tubes and were extracted using a modified version of the DNeasy® Blood and Tissue Kit (Qiagen). In each tube were added 380 μL of ATL Buffer and 20 μL of Proteinase K (Roche). The samples were placed in a theromixer for 1 hour at 56°C. The lysate was transferred in a different tube and 400 μL of AL Buffer, and 400 μL of 96% Ethanol were added. The extraction reaction was then spun down in a DNeasy Mini column, and the filter was washed with 500 μL of Buffer AW1 and 500 μL of Buffer AW2. The DNA was eluted twice using 50 μL of AE Buffer directly onto the DNeasy Mini column membrane.

Modern Italian wolves tissue samples - DNA sample collection, storage and extraction

Four Italian wolf samples were collected around 2010 and 2012 from road-killed individuals populating the Simbruini Mountain Range Regional Park and National Park of Abruzzo in the Central Apennines (See Table S2). The muscle samples were stored in ethanol at -20°C and subsequently processed in the “Modern
DNA laboratory” at the Fondazione Edmund Mach (FEM). Small pieces of tissue of around 25 mg were extracted using the DNeasy Blood and Tissue Kit (Qiagen) with overnight digestion at 56°C. The elution was conducted at the GLOBE Institute (University of Copenhagen) using two washes of 50 μL of AE buffer, with 10 minutes of incubation at 37°C. Until the elution, samples were stored at -20°C inside the DNeasy Mini spin columns.

Modern wolves - DNA sample collection, storage and extraction
Twenty-nine modern wolf (tissue or blood) samples from several locations in Europe (Table S2) were extracted using the Thermo Scientific KingFisher instrument and following the manufacturer’s protocol. The concentration of the samples was measured with a Qubit Fluorometer (ng/μL). All the extracts, with the exception of the modern Italian wolves and the Cirneco dell’Etna, were sent to BGI Copenhagen for library build and sequenced on 1/8 of a lane each on DNBSEQ PE150.

Library build, amplification and sequencing of modern Italian wolves and Cirneco dogs
Extracts were fragmented in the Covaris LE220 plus Focused-ultrasonicator with the parameters set for getting 350-bp fragment length. The extracts were diluted to obtain 100 ng concentration and BGI libraries for the Italian wolves and Cirneco dogs were constructed following Carøe et al. (2018)88 and Mak et al. (2017)64 using 10 μM adaptors. Libraries were purified using MinElute columns using PE buffer (Qiagen) and eluted in 60 μL of EB buffer.

The PCR mixture for the Italian wolf libraries consisted of: 20 μL of purified library, 0.2 μM of forward and reverse BGI primers, 2.5 μL of PfuTurbo Cx HotStart DNA Polymerase, 10 μL of Buffer 10X, 0.08 mg/mL BSA, 0.5 mM of dNTPs (25 μM) and 61.2 μL AccuGene molecular biology water (Lonza, Basel, CH). The Cirneco dog libraries were amplified using 20 μL of purified library, 0.2 μM of forward and reverse BGI primers, 0.05 μL of AmpliTaq Gold (Thermo Fisher Scientific, USA), 0.4 mg/mL BSA, 0.2 mM of dNTPs (25 μM), 10 μL of Buffer 10X, 2.5 mM of MgCl2 and 50.2 μL AccuGene molecular biology water in a total volume of 100 μL reaction.

The amplification of the Italian wolves was performed as follows: initial denaturation at 95°C for 2 minutes followed by 10 to 12 cycles of 95°C for 30 seconds, 60°C for 30 seconds, and 72°C for 110 seconds, and a final elongation step at 72°C for 10 minutes. PCR cycling conditions for Cirneco dog libraries were: initial denaturation at 95°C for 10 minutes followed by 12 cycles of 95°C for 20 seconds, 60°C for 30 seconds, and 72°C for 45 seconds, and a final elongation step at 72°C for 5 minutes. The Italian wolves and Cirneco dogs were sequenced on 1/8 of a lane each on MGiseq2000 PE150 and DNBSEQ PE150, respectively.

QUANTIFICATION AND STATISTICAL ANALYSIS

Data processing and historical DNA authentication
Short sequencing reads from modern, historical, and ancient samples were mapped to the dog reference genome (CanFam3.1)62 using Paleomix v.1.2.13 pipeline.65 First, adapter sequences were removed using AdapterRemoval 2.066 with default settings, and BWA v0.7.12 backtrack67 was used to perform the alignment of reads to the dog genome (bwa seed was disabled) setting the minimum base mapping quality to 0 to retain all the reads in this step. In further computational steps, reads with mapping quality lower than 20 or 30 (as specified in each analysis) were discarded. Picard MarkDuplicates v2.9.190 was used to identify and remove PCR duplicates and GATK v4.1.091,68 was used to perform the indel-realignment step. For the historical Sicilian samples in this study the post-mortem DNA profiles and misincorporation patterns were estimated through mapDamage2.072 (Figure S1).

Sex determination of the Sicilian wolves
We determined the genetic sex of four Sicilian wolves, by comparing the average genomic coverage with the coverage of the X chromosome. The samples Sic1, Sic2 and Sic3 were identified as males (XY) while Sic7 was shown to be female (XX) (Table S1).

Mitochondrial phylogeny
We investigated maternal lineage of the Sicilian wolves generating a mitochondrial phylogeny using RAxML-ng33 including all the wolves and dogs in our dataset and using the coyote as the outgroup. The Maximum-Likelihood mitochondrial phylogeny is consistent with previous publications15,16 and four out of
the five Sicilian wolves (Sic1, Sic2, Sic3 and Sic7) cluster together in the phylogeny as a sister clade to wolves from Romania and Slovakia (Figure S2D). They also fall within the same clade as wolves from Italy, Bulgaria, and Poland. The last sample, Sic4, falls within the dog mitochondrial diversity, as shown in a previous work based on the d-loop region. Unfortunately, we did not recover sufficient nuclear data from Sic4 to further investigate its ancestry. However, the sample was labelled as Canis lupus, but morphological analysis identified dog features, suggesting a hybrid origin. Our mitochondrial phylogenetic analysis placing Sic4 within dogs, is in agreement with the hybrid origin or, alternatively, it might suggest a potential mislabelling.

**Dataset**

The dataset compiled for this study is represented by 146 samples, including one Andean fox (Lycalopex culpaeus), 4 coyotes (Canis latrans), 62 modern wolves (Canis lupus) from Eurasia and North America, 45 modern dogs (Canis lupus familiaris), 24 ancient dogs (Canis lupus familiaris), and 5 Pleistocene wolves and 5 newly sequenced Sicilian wolves (Canis lupus cristaldi) (Table S2). In this dataset, 33 modern wolves and 3 Cirneco dogs were newly sequenced as part of this study.

**Variant calling**

In order to account for the varying depth of coverage of the samples and to incorporate the low-coverage ancient samples, we used a pseudo-haploid dataset instead of performing genotype calling. For each modern sample at each genomic site, we sampled a random read using ANGSD v0.93169 (-doHaploCall 1) from the reads with a minimum mapping quality of 30 (-minMapQ 30) and bases with a minimum quality of 20 (-minQ 20). Additionally, the following parameters were used: -minMinor 1 -maxMis 10 -skipTriallelic 1 -doMajorMinor 2 -C 50 -baq 1 -remove_bads 1 -only_proper_pairs 1 -rmTrans 1. Transitions were discarded (-rmTrans 1) to reduce the aDNA-derived error in the historical samples included in our dataset. The output file was converted into Plink files using the haploToPlink tool from ANGSD. Once the SNPs were chosen based on the modern genomes, we used ANGSD with -doHaploCall (option -s) to obtain pseudo-haploid calls for the ancient samples, historical samples, and outgroups (Coyote and Andean fox). The pseudo-haploid calls for the modern, ancient, and historical genomes were merged using Plink. The resulting dataset consisted of 74.5 million SNPs.

Plink v1.9.74 was used to further filter the pseudo-haploid panel. We kept sites with a minimum allele frequency of 5% (-maf 0.05) and minimum genotyping rates of 25% (-geno 0.25). Finally, we performed a linkage disequilibrium (LD) pruning in windows of size 10kb (-indep-pairwise 10kb 2 0.5). After filtering, the pseudo-haploid panel consisted of 2.3 million. This dataset was used in the following analyses: ADMIXTURE, MDS, f3-statistics, D-statistics, TreeMix and f4-statistics-based admixture graphs. From here on, we refer to this dataset as “pseudo-haploid SNP panel”.

**MDS**

The pseudo-haploid SNP panel described in the previous section (146 samples and a total of 2.3 million transversion sites) and Plink v1.9 –mds-plot74 option were used to perform a multidimensional scaling (MDS) analysis. We used Plink v1.9 –mds-plot option on two subsets of the samples: a subset including all modern and ancient wolf and dog samples (Figure 1A) and a second subset restricting to only modern and ancient wolves (Figure S2A).

**Admixture**

The pseudo-haploid SNP panel, restricted to modern and ancient wolves and dogs, was used to run ADMIXTURE v1.935 in order to estimate their ancestry components. The outgroups were not included in this analysis. ADMIXTURE was run assuming 2 to 20 ancestry components (K2 to K20). We ran 50 independent replicates for each value of K keeping the results from the replicate with the best likelihood. To explore which K best fitted our data, we performed cross-validation (-cv) for the best replicate for each K (Figure S2C). The results from the cross-validation procedure showed K=2 is the K that best fits our dataset. Therefore, we show the results from K=2 in Figure 1. Additionally, we show the results obtained with K=5 to show that the Sicilian wolf clusters with the Italian wolf, which is consistent with MDS and f2-statistic analyses (Figures 1A and 2A). Pong94 was used to visualise the admixture plots.
Mitochondrial and nuclear phylogenies

We built a mitochondrial phylogeny including 131 the samples in the dataset and using the coyote as the outgroup. First, we extracted the reads mapping to the chrMT from the genomic alignments using samtools. Then, we built a majority-count consensus sequence using ANGSD (-dofasta 2). The mitochondrial sequences for each sample were combined and aligned using MUSCLE. Finally, a ML tree was built using RAxML-ng, setting the evolutionary model to GTR+G and running 500 bootstrap replicates. The resulting tree was visualised using FigTree v.1.4.4.

To build the nuclear phylogeny on all the modern and ancient individuals in the dataset, we first used ANGSD to generate a random-read sequence for each genome using the dog reference genome (CanFam3.1) as reference and sampling a random reads at each genomic site (-dofasta1). Bases with base quality lower than 20 and reads with mapping quality lower than 20 were discarded (-minQ 20 -minmapq 20). The minimum coverage for each individual was set to 3x (-setminDepthInd 3), and the following additional filters were used: -doCounts 1 -remove_bads 1 -uniqueOnly 1 -baq 1 -C 50. Then, from autosomal chromosomes, we selected 1000 random regions, each 5000bp long, from the dog reference genome using BEDTools random and the following parameters: -l 5000 -n 1000. We used samtools to select from each sample the regions of interest and combined them into a multi-sequence alignment (MSA) fasta file. Each MSA file was used in IQ-TREE v.2.1.2 to reconstruct a phylogeny using 1000 bootstrap replicates (-B 1000) with UFBoot2, 1000 bootstrap replicates for SH-aLRT (-alrt 1000) and ModelFinder Plus to identify the best evolutionary model for each region. The resulting 1000 trees were concatenated using Astral-III to estimate the nuclear phylogeny. FigTree v.1.4.4 tool was used to visualise the species tree estimated with Astral-III.

Outgroup f3-statistics

We calculated outgroup f3-statistics using the qp3pop tool from Admixtools v.5.1 and the pseudo-haploid SNP panel to estimate the shared genetic drift between pairs of samples since their separation from an outgroup. For each of the Sicilian wolf genomes, we estimated an f3-statistic of the form f3(Andean Fox: Sicilian wolf, X), where X represents all remaining samples in the dataset (Figures 2A and S3). The more recently a pair of samples split the larger is the expected f3-statistic.

D-statistics

We computed D-statistics using the qpDstats tool from Admixtools v.5.1 and the pseudo-haploid SNP panel, to study how modern and ancient wolves and dogs are related to the Sicilian wolves. We estimated the following D-statistics tests:

\[ D(\text{Sicilian wolf, Italian wolf; X, Andean fox}) \]

Our treemix admixture graphs showed the most likely position in the tree for the Sicilian wolf is forming a clade with the Italian wolf. Therefore, we used this D-statistic to test if the Sicilian wolf shared more alleles with dogs (X) in comparison to the Italian wolf, which would be indicative of gene-flow. We performed this test for each of the four Sicilian wolf genomes independently (Figures 2B and S4A). Additionally, this test allowed us to determine the direction of the gene flow between dogs and the Sicilian wolf. If the gene flow was from the dogs into the Sicilian wolf, we would expect significant results regardless of the dog in X. Conversely, if the gene flow was from the Sicilian wolf into specific dog breeds we would expect significant results only when those breeds are placed in X. We find significant positive results for all dogs suggesting there is dog ancestry in the Sicilian wolf (Figure S4E).

\[ D(\text{Sicilian wolf, Portuguese wolf; X, Andean fox}) \]

We confirmed the result obtained in the previous test changing the Italian wolf for the Portuguese wolf (Figure S4E).

\[ D(X, Cirneco dog; Sicilian wolf, Andean fox) \] and \[ D(X, Croatian Eneolithic dog (ALP01); Sicilian wolf, Andean fox) \]

Where X corresponds to all dogs in the dataset. We performed these two tests in order to identify which of the Cirneco dogs and the Croatian Eneolithic dog (ALP01) was a better source for the dog ancestry in the Sicilian wolf. For every X, we expect the D-statistic to be larger for the dog that is a better source of ancestry.

[Reference]
(Cirneco or Croatian Eneolitic dog). Overall, we obtain larger values of D when the Croatian Eneolitic dog is used in the test (Figure S4B), confirming the Croatian Eneolitic dog is a better source of dog ancestry.

In all the D-statistic tests we used the Andean fox as the outgroup. We considered a test to be statistically significant if the resulting z-score was smaller than -3.3 or larger than 3.3. Finally, following Admixtools’s setup, in a test of the form D(H1, H2; H3, Outgroup), positive D-statistics indicate excess shared derived alleles between H1 and H3, while negative D-statistics indicate excess shared derived alleles between H2 and H3.

**TreeMix**

TreeMix was used to model the phylogenetic history of the Sicilian wolf and explore potential admixture events from dogs into Sicilian wolves. TreeMix was run on a subset of the samples including the four Sicilian wolves, Italian wolves, Iberian and other European wolves and ancient and modern dogs (a total of 37 samples). Starting from the pseudo-haploid SNP panel, we further restricted the analysis to sites without missing data resulting in a total of 160,233 transversion sites. We ran ten replicates for each migration event (-m), and incorporated from 0 to 4 migrations. The optional parameters -noss, -global and -k 500 were used and the tree was rooted using one of the coyote samples (Alabama coyote). The trees with the best likelihood for each number of migrations were selected using R97 and plotted using the R functions provided as part of the Treemix software. After incorporating the 1st migration (which models the Sicilian wolf as a mixture of wolf and dog) the topology of the clades involving the Sicilian wolf and Cirneco dog remains overall constant from up to 4 migration events, and none of the following inferred migrations involve the Cirneco dog or the Sicilian wolf.

**Admixture graphs**

To model the relationships between the Sicilian wolf, modern wolves and dogs we estimated an admixture graph using qpGraph. We included samples from relevant groups: Italian wolf (MW303, MW307, ItalianWolf_W050, ItalianWolf_W040), Iberian wolf (PortugueseWolf, MW122, MW127), modern dog (BoxerDog, GalgoDog, GShepDog), ancient dog (ALPO01_Croatia_Eneolithic, SOTN01_Croatia_Eneolithic, AL2397_Italy_Early), Sicilian wolf (Sic1, Sic7, Sic2) and the Andean fox. First, we identified samples that could be grouped into each of the relevant groups using the qpWave tool from AdmixTools. By comparing two lists of populations (left and right populations) qpWave identifies the minimum number of migrations required from the right populations to define the ancestry of the left populations. As ‘left populations’ we tested all possible pairs of Italian wolves, Iberian wolves, modern European dogs, ancient dogs and Sicilian wolves. For the ‘right populations’ we used all other samples, except the ones included among the ‘left populations’ and using the Andean fox as fixed outgroup. The qpWave tool was run using the ‘allsnps=YES’ option. Pairs of samples that were consistent with a single migration (p-value \( \leq 0.05 \)) were grouped into a population for the admixture graphs: Italian wolf (MW303, MW307 and ItalianWolf_W040), Iberian wolf (PortugueseWolf, MW122, MW127), ancient dog (ALPO01_Croatia_Eneolithic and AL2397_Italy_Early), modern European dog (GalgoDog and GShepDog) and Sicilian wolf (Sic1 and Sic7).

To efficiently explore the possible admixture graph models we used two approaches: qpBrute and a ‘base graph’ approach previously described. For the qpBrute approach we created a configuration file for qpBrute and performed a heuristic search of the admixture graph space. For the ‘base graph’ approach we use admixturegraph R library to list all possible graphs including the Italian wolf, Iberian wolf, ancient dog, and modern European dog, and fixed the Andean fox as an outgroup. We tested all possible graphs including 0, 1 and 2 migration edges. We then included the Sicilian wolf to the two graphs that fitted our data in terms of the z-score of the worst fitting f4-statistic and with a score that was not significantly different. The Sicilian wolf was included both as a non-admixed lineage and as an admixed lineage. In Figures S4C and S4D we show the best graph from each approach from the graphs that fitted our data.

**PCAdmix**

PCAdmix was used to infer local ancestry for the Sicilian wolves. We used 2 Cane Corso dogs, 3 Cirneco dogs, and 4 European dogs (Galgo, German shepherd, boxer, Tibetan mastiff) as representatives of the dog ancestry, and 4 Italian wolves and 6 European wolves as representatives of the wolf ancestry.

An imputed and phased dataset was prepared for PCAdmix analysis. Genotype likelihoods were called using samtools for 139 canid individuals in our dataset and then used as input for genotype imputation.
without a reference panel using Beagle v4.1. The imputed dataset was then phased using Beagle v5.4. We used the recombination map of the domestic dog genome for imputation and phasing.

PCAdmix was run with a window size of 0.01 cM (-wcM 0.01) and the input SNPs were pruned by minor allele frequency (-maf 0.1) and LD (-r2 0.8). We used a threshold of 0.9 for posterior probability to infer the ancestry and assigned an “unclear” ancestry for regions below this threshold. In order to evaluate the performance of the inference, we also inferred ancestry for randomly selected dogs and wolves including one wolf (wSierraMorena) with known admixed ancestry.

Nucleotide diversity and heterozygosity in sliding windows

The genome-wide nucleotide diversity (\(p\)) was estimated for four wolf populations: 4 Sicilian wolves, 5 Italian wolves, 4 Scandinavian wolves and 6 Iberian wolves. Vcftools 0.1.16 was run on the pseudo-haploid SNPs panel with a window of 100kbp. Rstudio and ggplot were used to visualise the data for each population.

We used ANGSD to estimate the heterozygosity of each sample, by calculating the folded site frequency spectrum (SFS) on the autosomal chromosomes. We estimated genotype likelihoods for each of the samples independently using ANGSD’s GATK (-GL 1) model (doSaf 1 -fold 1), removing transitions (-rmtrans 1), bases with quality score lower than 20 (-minQ 20) and reads with mapping quality lower than 20 (-minmapq 20). The dog reference genome was used both as reference and as ancestral (-ref and -anc options) and the repeat regions were masked using a repeat mask file (-sites). The chromosomes were partitioned into overlapping windows of size 1 Mb with a step size of 500 kb using the BEDTools windows tool. Windows shorter than 1Mb at the end of the chromosomes were discarded. The SFS for each window was estimated using the realSFS utility tool provided in ANGSD and subsequently the final heterozygosity per window was calculated as the ratio of heterozygous sites/total sites. Rstudio, ggplot2 and dplyr were used to visualise the heterozygosity level at each window in the form of a violin plot for each individual.

ROHan

To assess the extent of recent inbreeding, we estimated the length and abundance of segments of runs of homozygosity (ROH) using ROHans. We estimated ROHs on two genomes of Sicilian wolf (Sic1 and Sic2) with ~5x (the acceptable minimum coverage for ancient genomes when post-mortem deamination is <=0.01), 2 Pleistocene wolves with minimum coverage of 7x, and 11 modern wolves with minimum coverage of 5x. Before running ROHan, we created a bed file containing mappable regions in the reference dog genome (CanFam3.1) to account for divergence between the genomes of the wolf and the dog. Then, a file listing only the autosomal chromosomes was created to restrict our analysis to the autosomal regions. To obtain an estimate of the expected number of segregating sites in the ROH segments (~rohmu) required by ROHans, we estimated background heterozygosity using the Watterson’s theta formula implemented in ROHans. In this formula, the number of segregating sites is calculated as four times the mutation rate multiplied by the effective population size. We consider that the effective population size of contemporary grey wolf populations is ~1,000 following (Pacheco et al., 2022) and a mutation rate of 1x10^-8, resulting in an expected number of 4x10^-4 segregating sites, which means that in a ROH segment it will be lower at 2x10^-4. We also observed the deamination profile of the ancient genomes using bam2prof (included in ROHans) and determined that for most of the ancient genomes the damage levels were off at -q 20. The bam2prof results were then used on the ROHans call for the Sicilian wolf genomes (~deam5p and ~deam3p).

We calculated the number of segments (NROH), mean length (LROH), and total sum of runs of homozygosity (SROH) in the autosomal region of each sample from the .hmmroh output of ROHans using an R script. The genomic coefficient of inbreeding was calculated in the same script by dividing the SROH by the total number of validated sites by ROHans available on the .hEst file. The plots in Figures 3C and 3D show the middle mid-point estimates from the three mid-point estimates provided by ROHans (min, mid, and max).

Transversion load

To assess the mutational load in the two Sicilian and 13 reference wolf genomes where we calculated ROH, we used the SIFT scores as measure of mutation’s deleteriousness based on the change it causes in the protein, which is available for the domestic dog, Canis familiaris (CanFam3.1), through Variant Effect Predictor (VEP). We first extracted positions of coding sequences (CDs) from the domestic dog genome.
available from the GFF file from Ensembl (https://www.ensembl.org/info/genome/variation/prediction/predicted_data.html#consequences) (release 104; accessed: 15 July 2022). The VEP was used to get SIFT scores at the positions at the CDs, limiting to sites marked as canonical (–pick 1) with scores smaller than 0.05 to get the most deleterious positions. As SIFT scores differ according to base, VEP was run for all four possible variants (A, C, G, or T). This SIFT reference file was then run with a custom script that weights the scores based on the genotype probability in that variant’s position (see below for how genotype probabilities were estimated). The mutational load was calculated in this set of predetermined sites that were attainable from all compared samples (a total of 11,021,945 sites that were present in all 15 samples compared in Figure 3).

As SIFT scores are the normalised probability that an amino acid change is tolerated, with scores lower than 0.05 predicted as deleterious,\textsuperscript{106} the magnitude of SIFT scores is negatively correlated with the deleteriousness of a mutation (i.e. the smaller the score, the more deleterious the mutation is). To generate a more intuitive metric that can be used to compare the deleteriousness across samples, we estimated the SIFT complement score ($C_d$) by subtracting 1 from the SIFT score for each derived allele $d$:

$$C_d = 1 - \text{SIFT}_d$$

We then estimate the mutational load at each site $i$, by multiplying $C_d$ by the genotype probability of allele $d$ at site $i$. For each site (i) where the ancestral allele (a) is known, three different derived alleles ($d$) are possible given that $d \neq a$ and $d,a \in \{A,C,G,T\}$. Each of these derived alleles have their own complement score ($C_{ad}$, $C_{gd}$ and $C_{gd}$). The homozygous derived load for each site $i$ is thus:

$$S_i(ho_d) = \sum_{k=1}^3 P_i(G_k dik) \cdot C_{dk}$$

(Equation 1)

where $P_i(G dik)$ is the genotype probability of genotype $G dik$ that is homozygous for allele $d$ in position $i$. Thus, for a genome with a total of $n$ positions, the total sum of conservation scores in homozygous positions ($S_{HOD}$) is simply the sum of homozygous derived score (Equation 1) across sites:

$$S_{HOD} = \sum_{i=1}^n S_i(ho_d)$$

(Equation 2)

The entire homozygous load ($L_{HOD}$) for a genome with $n$ sites is thus:

$$L_{HOD} = \frac{S_{HOD}}{\sum_{i=1}^n \sum_{k=1}^3 P_i(G dik)}$$

(Equation 3)

When dealing with ancient genomes, removing transitions is desirable to exclude errors derived from post-mortem DNA damage. In other words, homozygous transversion load is calculated using Equation 1 for $d \in \{C,T\}$ only when $a \in \{A,G\}$ and for $d \in \{A,G\}$ only when $a \in \{C,T\}$. As SIFT scores are different for different alleles, the homozygous transversion load ($L_{HOTV}$) for a genome with $n$ sites is:

$$L_{HOTV} = \frac{\sum_{i=1}^n P_i(G dik) \cdot C_{dik} + P_i(G dik) \cdot C_{dik}}{\sum_{i=1}^n \sum_{k=1}^3 P_i(G dik)}$$

(Equation 4)

where $P_i(G dik)$ is the genotype probability for homozygous transversion site 1 and $C_{dik}$ is its respective conservation score, $P_i(G dik)$ is the genotype probability for homozygous transversion site 2 and $C_{dik}$ is its respective conservation score, and $P_i(G dik)$ is the probability of homozygous transversion genotype.

Estimating genotype probabilities used in the estimation of the homozygous transversion load

The genotype probability was obtained from the genotype likelihood values obtained from ANGSD\textsuperscript{69} with minimum base quality 20 (-minQ 20), minimum mapping quality 20 (-minMapQ 20), 5 edge bases trimmed (-trim 5), genotype likelihood estimation following GATK model (-GL 2) to have an output the log likelihoods of all 10 genotypes (-doGlf 4). A framework employing a Bayesian framework described in Plassais et al. (2022)\textsuperscript{107} was used to convert the genotype likelihood to a genotype probability using a custom script made available in a GitHub repository (Greer, 2020,\textsuperscript{51} all_genotype_probability_v3.py).

Estimating the ancestral allele

As the above equations require the ancestral state of each position, we inferred the ancestral allele by mapping five species representing taxa that are ancestral to the genus Canidae to enable the comparison of all
taxa within the Canidae family. These ancestral taxa were used to polarise the alleles and included: domestic cat (Felis catus), polar bear (Ursus maritimus), spotted hyena (Crocuta crocuta), red panda (Ailurops fulgens), and walrus (Odobenus rosmarus). These taxa were inferred to be ancestral to the Canidae group using the mammal tree of life.\textsuperscript{108} The domestic cat (Felis_catus_9.0, GCA_000181335.4) and polar bear (Urs-Mar_1.0, GCA_000687225.1) fasta files were downloaded from Ensembl, while the hyena, red panda, and walrus reference genomes were generated by the DNA Zoo consortium following the Juice Box protocol.\textsuperscript{109,110} The walrus and red panda assemblies were generated based on available draft genomes.\textsuperscript{111,112} The reference-quality genomes of each of the taxa, available as a fasta file, were split into 500bp fragments and aligned to the CanFam3.1 reference genome using bwa-mem.\textsuperscript{113} The resulting BAM files were merged and a consensus sequence was called using ANGSD -doFasta 2.