

Electronic Supplementary Materials

Genetic footprint of population fragmentation and contemporary collapse in a freshwater cetacean

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ESM Text S1. Supplementary details on the ABC analyses

The parameters defining each scenario (i.e. population sizes, times of population size changes and splits, and mutation rates) were considered as random variables drawn from prior distributions (figure 2, ESM table S1, S2). For each simulation, DIYABC draw a value for each parameter from its prior distribution and performed coalescent simulations to generate a simulated POD with the same number of gene copies and loci per population as in the observed dataset. A set of 48 summary statistics (SS) describing within and among population genetic diversity were calculated with DIYABC for each POD and the observed data. Within population statistics for microsatellite loci included the mean number of alleles per locus (NAL), expected heterozygosity (HET), allele size variance (VAR), M_{GW} statistic of Garza & Williamson across loci [1]. Between population statistics for microsatellites included F_{ST} [2], shared allele distance (DAS) [3], and $(\delta\mu)^2$ Goldstein's distance [1,4]. For the mtDNA data, the descriptive statistics within populations include the number of haplotype (NAH), the number of segregating sites (NSS), the mean pairwise differences (MPD) and its variance (VPD), Tajima's D , and the number of private segregating sites (PSS). Statistics computed between groups were the mean of within sample pairwise differences (MP2), mean of between sample pairwise differences (MPB), and H_{ST} between two samples [2,5]. A Euclidean distance was calculated between the statistics obtained for each normalized PODs and observed dataset [3,6].

Coalescent simulations assume a mutation model for each type of loci. The mutation model for microsatellite loci was a generalized stepwise-mutation (GSM) model [7] with two parameters: a mean mutation rate ($\bar{\mu}_{mic}$) and mean of the geometric distribution for the length, in repeat numbers, of mutation events (\bar{P}) drawn from uniform prior distributions ($\bar{\mu}_{mic} : [10^{-3} - 10^{-4}]$ and $\bar{P} : [0.1-0.3]$, see Table S2). We accounted for variation in μ_{mic} and P among loci by drawing their individual values from a gamma distribution (Table S2). These settings allowed for large mutation rate variance across loci (i.e. range of 10^{-5} to 10^{-2}). We also considered mutations inserting or deleting a single nucleotide in the microsatellite sequence.

We used jModelTest 2.1.7 [8] to identify the best substitution model describing the sequence variation of the mtDNA-CR and estimate its parameter. The best mutation model was a HKY+I+G model [9] with a proportion of constant sites of 16%, and a shape of the gamma

distribution of mutations among sites equal to 99.8 (Table S1). We assumed a per-site and per-generation mutation rate ranging uniformly between 1×10^{-7} and 1×10^{-5} , as found in the literature [10,11].

The PPr of each competing scenario was estimated using a polychotomous logistic regression [12,13] on the 1% of simulated datasets closest to the observed dataset (lowest Euclidean distance δ), subject to a linear discriminant analysis as a pre-processing step (to reduce the dimensionality of the data) [14]. The best-fitting scenario was selected based on the highest PPr value with a non-overlapping 95% confidence interval (95% CI). We estimated the posterior distributions of each demographic parameter under the best demographic model, by carrying out local linear regressions on the 1% closest of 10^6 simulated data sets, after a logit transformation to parameter values [6,12].

We evaluated the ability of the ABC analysis to discriminate between the competing scenarios by analysing 300 simulated data sets with the same number of loci and individuals as our real data set. Following Cornuet *et al.* [13], we estimated the Type-I error rate as the proportion of instances in which the selected scenario did not show the highest posterior probability among the competing scenarios, for the 300 simulated datasets generated under the best-supported scenario. Similarly, we estimated the Type-II error rate, by simulating 300 data sets for each of the other competing scenarios and calculating the mean proportion of instances in which the best-supported model was incorrectly selected as the most probable scenario. Finally we conducted a model checking procedure implemented in DIYABC to evaluate the goodness-of-fit between the posterior parameter distribution and the observed data following [15]. For this analysis, we simulated 1,000 pseudo-observed datasets under each model-posterior combination, with sets of parameter values drawn with replacement from the parameter posterior distribution. This generated a posterior cumulative distribution function for each summary statistic allowing us to estimate how well each fitted model can reproduce the observed summary statistics.

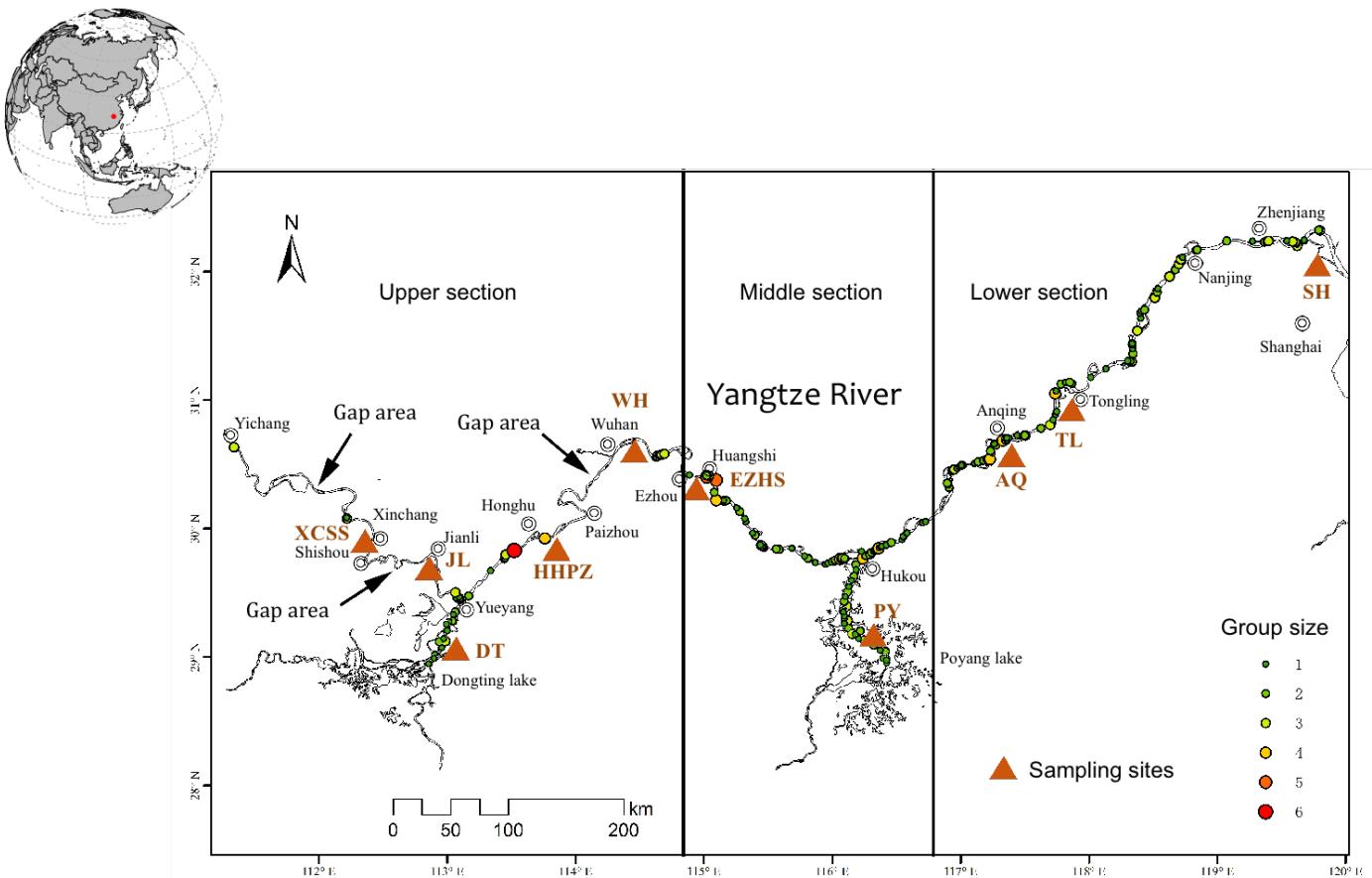


Figure S1. Maps showing the distribution of the Yangtze finless porpoises in the Yangtze River and sampling sites. Orange triangles show the locations of each sampling site and their acronyms. Sightings and distribution gaps reported during field surveys [16-18] are shown with color-coded dots and arrows, respectively. The top-left insert show the location of the Poyang Lake (red dot) on globe.

Step 1

- **Coalescent simulations under each scenario (SC)**

- Simulation of 10^6 pseudo-observed data sets (PODs) for each SC.
 - Parameter values (N, t, μ) drawn with replacement from prior parameter distributions
 - Calculate summary statistics for each POD (SS) and for the observed real data set (SS*)

Step 2

- **Model selection: Posterior probability (Pp) of each competing scenario**

- Polychotomous logistic regression on 1% PODs producing the closest SS value to the observed SS*, after a linear discriminant analysis as a pre-processing step.
- Identification of the best SC based on non-overlapping Pp 95% confidence intervals

Step 3

- **Confidence in scenario choice**

- Simulate 500 test data sets with each SC.
 - For each test dataset, estimate Pp as in STEP2.
 - Record error rates produced by each SC to estimate Type-I and Type-II error rates and power to discriminate SC.

Step 4

- **Estimation of marginal parameter posterior distributions**

- Local linear regressions on the closest 1% of 10^6 simulated data sets, after the application of a logit transformation to parameter values.

Step 5

- **Model checking** (assessment of the goodness-of-fit of a model parameter posterior combination)

- Simulating 1,000 pseudo-observed data sets under each studied model-posterior distribution combination.
- Sets of parameter values drawn with replacement from the 1,000 sets of the posterior sample.
- Estimate the p-value of the deviation of the observed value of each statistic from its simulated distribution under the best demographic model.

Figure S2. Flow chart about inference of population history using ABC in the program DIYABC.

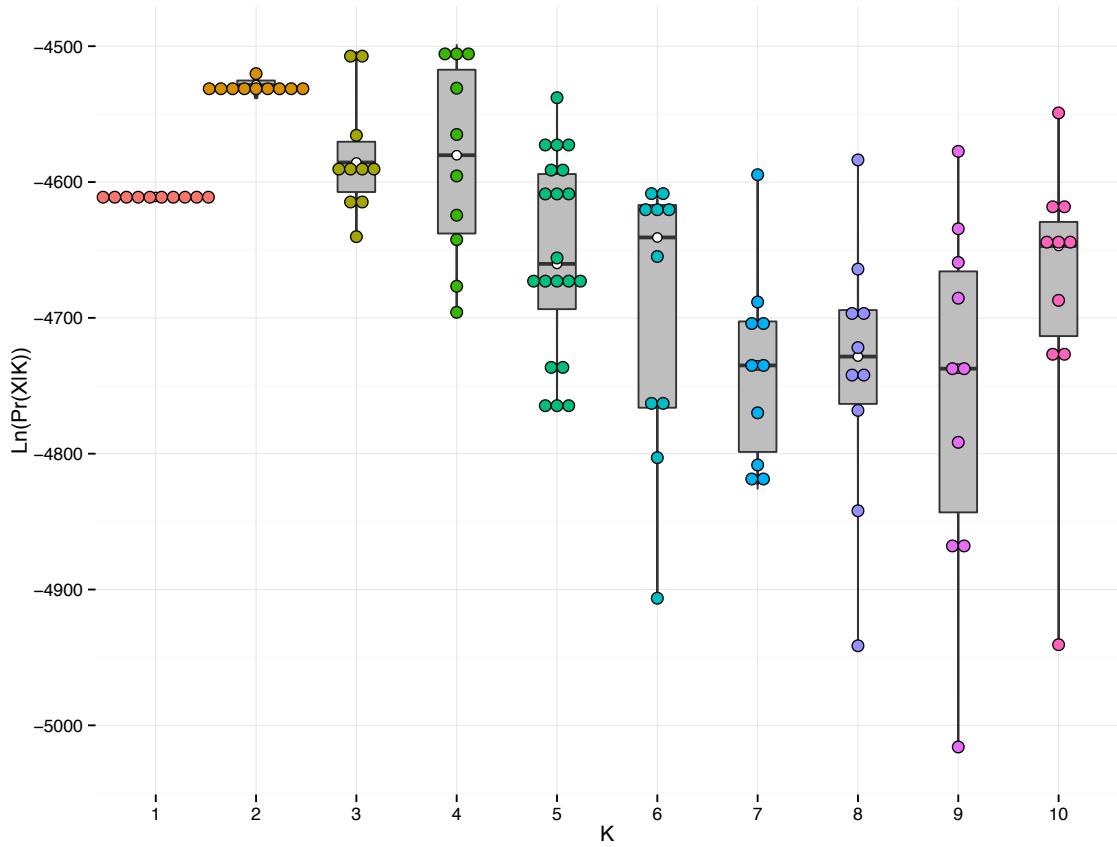


Figure S3. Estimated probability of the data (X) given K group tested in *Structure*.

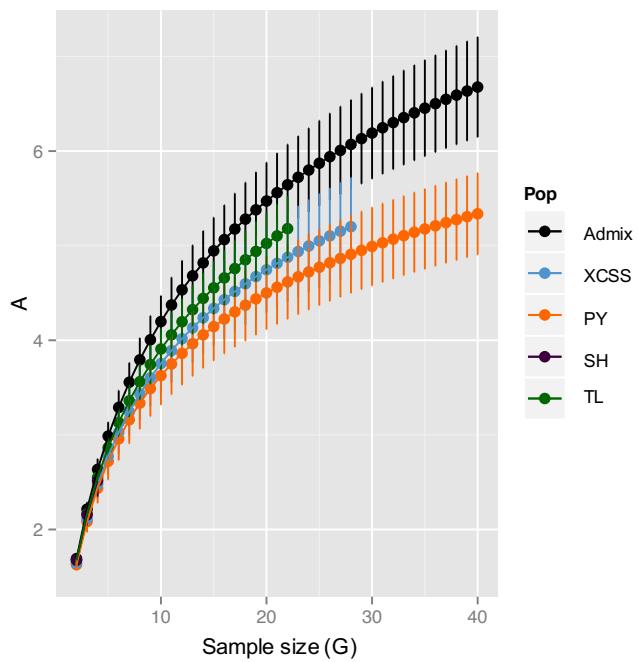


Figure S4. Allelic richness estimated using ADZE [19] for each group identified in *Structure*. Variation in Ar and pA are provided as a function of the sample size considered in the rarefaction procedure.

Table. S1: Model specification and prior distributions for demographic parameters.
See figure 2 for the demographic parameters of each model tested.

Step 2a.		
Demographic Parameter	Type	Prior
N_1	N	UN~[1–5,000]
N_2	N	UN~[1–5,000]
N_3	N	UN~[1–5,000]
N_a	N	UN~[10–20,000]
N_{TI}	N	UN~[10–10,000]
T_{rad}	T	UN~[10–5,000]
$T_1 (< T_2)$	T	UN~[10–10,000]
T_2	T	UN~[10–10,000]

Type of parameters: (N) effective population size, (T) time of the event in generation. Uniform distribution (UN) with 2 parameters: min and max; Gamma distribution (GA) with 4 parameters: min, max, mean and shape; Log-Uniform (LU) distribution with 2 parameters: min and max.

Table. S1 (Continue) - Step 2b_i

Demographic Parameter	Type	Prior
N11	N	UN~[1-5000]
N21	N	UN~[1-5000]
N31	N	UN~[1-5000]
Tanc1	T	UN~[10-5000]
Nanc1	N	UN~[10-20000]
N12 (>Nbot12)	N	UN~[1-5000]
N22 (>Nbot22)	N	UN~[1-5000]
N32 (>Nbot32)	N	UN~[1-5000]
Texp2	T	UN~[10-5000]
Nbot12	N	UN~[1-10]
Nbot22	N	UN~[1-10]
Nbot32	N	UN~[1-10]
db2	T	UN~[1-10]
Nanc2	N	UN~[10-20000]
N13 (<Nexp3)	N	UN~[1-5000]
N23 (<Nexp3)	N	UN~[1-5000]
N33 (<Nexp3)	N	UN~[1-5000]
Tisol3 (<Texp3)	T	UN~[10-5000]
Nexp3	N	UN~[11-10000]
Texp3	T	UN~[10-5000]
Nbot3 (<N13, N23, N33)	N	UN~[1-10]
db3	T	UN~[1-10]
Nanc3	N	UN~[10-20000]
N14	N	UN~[1-100]
N24	N	UN~[1-100]
N34	N	UN~[1-100]
Tcrash14	T	UN~[1-50]
Niso14 (<Nexp4)	N	UN~[101-5000]
Tcrash24	T	UN~[1-50]
Niso24 (<Nexp4)	N	UN~[101-5000]
Tcrash34	T	UN~[1-50]
Niso34 (<Nexp4)	N	UN~[101-5000]
Tisol4 (<Texp4)	T	UN~[51-5000]
Nexp4	N	UN~[101-10000]
Texp4	T	UN~[51-5000]
Nbot4	N	UN~[1-10]
db4	T	UN~[1-10]
Nanc4	N	UN~[10-20000]
N15 (<Nexp5)	N	UN~[1-5000]
N25 (<Nexp5)	N	UN~[1-5000]
N35 (<Nexp5)	N	UN~[1-5000]
Tisol5 (<Texp5)	T	UN~[10-5000]
Nexp5 (>Nfond5)	N	UN~[10-10000]
Texp5	T	UN~[10-5000]
Nfond5	N	UN~[1-10]
N16 (<Nexp6)	N	UN~[1-5000]
N26 (<Nexp6)	N	UN~[1-5000]
N36 (<Nexp6)	N	UN~[1-5000]
Tisol6 (<Texp6)	T	UN~[10-5000]
Nexp6	N	UN~[10-10000]
Texp6	T	UN~[10-5000]
Nanc6	N	UN~[10-5000]

Table S1 (Continue) – Step 2b_{ii}

Demographic Parameter	Type	Prior
N11(>Nbot1)	N	UN~[1, 5000]
N21(>Nbot1)	N	UN~[1, 5000]
N31(>Nbot1)	N	UN~[1, 5000]
Tisol1(<Texp1)	T	UN~[10, 5000]
Nexp1(>N11, N12,N13)	N	UN~[11, 10000]
Texp1	T	UN~[10, 5000]
Nbot1	N	UN~[1, 10]
db1	T	UN~[1, 10]
Nanc1	N	UN~[10, 20000]
N12	N	UN~[1, 100]
N22	N	UN~[1, 100]
N32	N	UN~[1, 100]
Tcrash12	T	UN~[1, 5]
Niso12(>N12)	N	UN~[1, 5000]
Tcrash22	T	UN~[1, 5]
Niso22(>N22)	N	UN~[1, 5000]
Tcrash23	T	UN~[1, 5]
Niso23(>N32)	N	UN~[1, 5000]
Tisol2(>Tcrash12, Tcrash22, Tcrash23)	T	UN~[10, 5000]
Nexp2	N	UN~[11, 10000]
Texp2(>Tisol2)	T	UN~[10, 5000]
Nbot2	N	UN~[1, 10]
db2	T	UN~[1, 10]
Nanc2	N	UN~[10, 20000]

Table S2. Mutation model parameters for the ABC analysis figure 2

Microsatellites mutation parameter	GSM (40 steps allowed)
$\bar{\mu}_{mic}$	UN~[$1 \times 10^{-4} - 1 \times 10^{-3}$]
$G_{\mu_{mic}}$	GA~[$1 \times 10^{-5}, 1 \times 10^{-2}, \bar{\mu}_{mic}, 2$]
\bar{P}	UN~[$1 \times 10^{-1}, 3 \times 10^{-1}$]
GP	GA~[$1 \times 10^{-2}, 9 \times 10^{-1}, \bar{P}, 2$]
SNI	LU~[$1 \times 10^{-8}, 1 \times 10^{-5}$]
G_{SNI}	GA~[$1 \times 10^{-9}, 1 \times 10^{-4}, SNI, 2$]
MtDNA mutation parameter	HKY (<i>p-inv</i> : 15.6, α : 99.8)
μ_{seq}	UN~[$1 \times 10^{-7}, 1 \times 10^{-5}$]
KI	UN~[0.050, 20]

The mutation model parameters for the microsatellite loci were the mean mutation rate (μ_{mic}), the parameter determining the shape of the gamma distribution of individual loci mutation rate (P), and the Single Insertion Nucleotide rate (SNI). The mtDNA mutation model was an HKY with two variable parameters, the per-site and generation mutation rate (μ_{seq}) and the transition/transversion ratio (KI) parameter, and two fixed parameters, the proportion of constant sites (*p-inv.*), and the shape of the Gamma distribution of mutations among sites (α).

Table. S3. Effective population size (N_e) estimated in each population of the Yangtze finless porpoise. Values have been calculated using an estimator based on linkage disequilibrium between loci in *NeEstime* [20], using an ABC approach in *ONESAMP* [21], and using DIYABC under SC2 Fig. 2*bii*.

	XCSS	TL	PY	Admix	Total
<i>LD–NeEstime</i> <i>Mean [95%CI]</i>	13 [7-25]	80 [23-∞]	89 [47-298]	106 [56-406]	288
<i>ONESAMP</i> <i>Mean [95%CI]</i>	22 [11-26]	18 [15-28]	62 [45-115]	80 [56-142]	182
<i>DIYABC</i> <i>Mode (mean) [95%CI]</i>	14 (42) [7 – 95]	32 (55) [11 – 98]	35 (50) [11 – 66]	-	-

Table. S4. Estimated recent migration (upper and lower matrix) and non-migration rates (along the diagonal) per generation between populations identified by *Structure*. Values are provided as geometric mean (median) [95% Highest probability density intervals].

To: From:\diagdown	Admix	PY	TL	XCSS
Admix	0.792 (0.791) [0.696 – 0.893]	0.011 (0.014) [0.000 – 0.045]	0.208 (0.238) [0.085 – 0.315]	0.276 (0.282) [0.222 – 0.326]
PY	0.185 (0.195) [0.095 – 0.292]	0.972 (0.975) [0.941 – 0.997]	0.056 (0.061) [0.000 – 0.211]	0.015 (0.019) [0.000 – 0.069]
TL	0.004 (0.005) [0.000 – 0.019]	0.003 (0.004) [0.000 – 0.016]	0.682 (0.678) [0.667 – 0.713]	0.009 (0.011) [0.000 – 0.045]
XCSS	0.004 (0.005) [0.000 – 0.022]	0.003 (0.004) [0.000 – 0.016]	0.009 (0.011) [0.000 – 0.046]	0.682 (0.678) [0.667 – 0.711]

(non-zero migration rate values are in bold font)

Table S5. Model selection procedure and performance analysis for the ABC step 1 (see figure 2a).

	SC1	SC2	SC3	SC4	SC5	SC6	SC7	SC8	SC9	SC10
Post. Prob.	66.7%	1.3%	1.4%	7.7%	5.8%	7.1%	2.8%	3.3%	2.4%	1.5%
95%CI	[65.6-67.8]	[1.2 - 1.5]	[1.2 – 1.6]	[7.1 – 8.2]	[5.3 – 6.2]	[6.6 – 7.7]	[2.5 – 3.0]	[3.0 – 3.6]	[2.1 – 2.6]	[1.3 – 1.7]
Performance analysis (based on 300 Pseudo Observed simulated Data sets, PODs)										
	SC1	SC2	SC3	SC4	SC5	SC6	SC7	SC8	SC9	SC10
D1	61.4%	11.6%†	18.3%†	14.3%†	13.7%†	12.0%†	9.0%†	14.0%†	11.8%†	11.7%†
D2	1.7%*	38.5%	17.6%	0.7%	1.8%	1.0%	0.3%	0.7%	0.0%	18.7%
D3	7.6%*	21.1%	30.0%	3.0%	2.4%	4.0%	3.0%	6.0%	3.0%	17.7%
D4	3.0%*	1.0%	1.3%	40.3%	24.0%	20.0%	1.0%	1.0%	1.6%	1.0%
D5	9.7%*	2.3%	4.3%	18.6%	33.4%	22.3%	4.3%	5.3%	6.0%	5.3%
D6	2.3%*	1.6%	1.0%	18.0%	17.1%	35.3%	1.0%	0.7%	0.7%	0.3%
D7	3.0%*	0.3%	1.0%	1.7%	1.0%	1.7%	34.0%	17.3%	23.3%	0.3%
D8	7.0%*	1.7%	4.7%	3.0%	5.5%	3.3%	23.0%	34.0%	18.0%	2.7%
D9	2.3%*	0.0%	0.7%	0.0%	0.8%	0.3%	23.0%	20.7%	35.3%	0.3%
D10	2.0%*	22.0%	21.0%	0.3%	0.3%	0.0%	1.3%	0.3%	0.3%	42.0%
Model check (number of outlying statistics, table S4)										
P < 0.05	1	3	7	9	4	8	2	4	4	1
P < 0.01	4	1	2	1	0	2	2	0	0	3
P < 0.001	0	0	0	0	0	0	0	0	0	1

D: proportion of case in which the simulation-based model choice procedure was able to select a scenario as the most probable with non-overlapping confidence intervals of the posterior probabilities of each scenario.

* Type-I or α -error rate; † Type-II or β -error rate and that $1 - \sum \beta_i$ provide the power of the model choice procedure.

Table S6. Model checking for the first step (see figure 2a).

OBS. VALUE	SC1	SC2	SC3	SC4	SC5	SC6	SC7	SC8	SC9	SC10	
MGW	0.5121	0.0055 (**)	0.0130	0.0105	0.0120	0.0115	0.0100	0.0090	0.0220	0.0115	0.0065 (*)
MGW	0.4598	0.0065 (**)	0.0175	0.01	0.0175	0.0160	0.0220	0.0110	0.0175	0.0130	0.0165 (*)
MGW	0.4403	0.006 (**)	0.0115	0.01	0.0100	0.0125	0.0060	0.0035	0.0150	0.0200	0.0060 (*)
DM2	5.0991	0.955 (*)	0.87	0.9095	0.9610	0.9570	0.9720	0.888	0.866	0.872	0.9115 (**)
NHA	1	0.053	0.0605	0.0475	0.0455	0.059	0.0485	0.0755	0.066	0.0825	0.0665
NSS	0	0.053	0.0605	0.0475	0.0455	0.059	0.0485	0.0755	0.066	0.0825	0.0665
MPD	0	0.053	0.0605	0.0475	0.0455	0.059	0.0485	0.0755	0.066	0.0825	0.0665
VPD	0	0.053	0.0605	0.0475	0.0455	0.059	0.0485	0.0755	0.066	0.0825	0.0665
MNS	0	0.053	0.0605	0.0475	0.0455	0.059	0.0485	0.0755	0.066	0.0825	0.0665
HST	0.7656	0.995 (**)	0.9950	0.986	0.9740	0.9565	0.9530	0.9570	0.9790	0.9640	0.9940 (*)
											(**)

Scenarios are shown in figure 2. The probability $Prob. (S_{simul} < S_{obs})$ for each summary statistic is calculated from 1,000 PODs simulated from the posterior distributions of parameters obtained under the each focal scenario. Corresponding tail-area probabilities (*p*-values) were obtained as $Prob. (S_{simul} < S_{obs})$ and $1.0 - Prob. (S_{simul} < S_{obs})$ for $Prob. (S_{simul} < S_{obs}) \leq 0.5$ and > 0.5 , respectively (*, **, *** = tail-area probability < 0.05 , < 0.01 and < 0.001 , respectively). In addition to the statistics used during the model choice procedure, the model check procedure used also two samples statistics including the mean number of alleles, mean genetic diversity, and mean size variance for microsatellite loci; and for mtDNA the within sample statistics including mean number of the rarest nucleotide at segregating sites and its variance, and the two samples statistics comprising the number of haplotypes and number of segregating sites. Only significant summary statistics for at least one scenario are shown. Abbreviations for the summary statistics are as follows: Mean Garcia-Williamson index (MGW), $d\mu^2$ distance (DM2); Number of mtDNA haplotypes (NHA); Number of segregating site (NSS); Mean pairwise difference (MPD); Variance of pairwise difference (VPD); Mean number of the rarest nucleotide at segregating sites (MSN), Hudson's FST-statistics (HST)

Table S7. Model selection procedure and performance analysis for the ABC step 2bi (figure 2bi).

	SC1	SC2	SC3	SC4	SC5	SC6
Post.Pr	6.10%	3.40%	77.80%	4.50%	7.20%	0.10%
95%CI	[5.6 – 6.7]	[3.0 – 3.9]	[76.8 – 78.9]	[4.0 – 4.9]	[6.7 - 7.7]	[0.0 – 0.1]
Performance analysis (based on 300 simulations)						
	SC1	SC2	SC3	SC4	SC5	SC6
D1	47.30%	17.30%	13.70%*	6.70%	7.30%	0.00%
D2	14.00%	60.00%	7.00%*	2.33%	4.70%	0.00%
D3	11.30%†	5.30%†	37.30%	6.30%†	13%†	0.30%†
D4	8.70%	9.30%	20.30%*	78.70%	34.00%	0.00%
D5	18.30%	6.30%	20.70%*	12.00%	41.00%	0.00%
D6	0.30%	1.70%	1.00%*	0.00%	0.00%	99.70%
Model check (number of outlying statistics, see table S6)						
P < 0.05	8	4	0	9	1	7
P < 0.01	2	1	1	0	4	0
P < 0.001	0	0	0	3	0	0

(see the legend of tabe S5)

Table S8: Model checking for the ABC step 2bi (figure 2bi).

	Obs. value	SC1	SC2	SC3	SC4	SC5	SC6
VAR_1_1	11.2626	0.686	0.68	0.43	0.9600 (*)	0.85	0.644
VAR_1_2	16.2651	0.833	0.836	0.618	0.9850 (*)	0.937	0.834
VAR_1_3	15.9953	0.837	0.823	0.627	0.9870 (*)	0.928	0.843
MGW_1_1	0.5121	0.0070 (**)	0.0130 (*)	0.0975	0.0000 (***)	0.0040 (**)	0.0350 (*)
MGW_1_2	0.4598	0.0135 (*)	0.0190 (*)	0.12	0.0010 (***)	0.0030 (**)	0.0365 (*)
MGW_1_3	0.4403	0.0115 (*)	0.0090 (**)	0.136	0.0010 (***)	0.0050 (**)	0.0375 (*)
V2P_1_1&2	13.0832	0.738	0.74	0.489	0.9710 (*)	0.882	0.725
V2P_1_1&3	12.4376	0.719	0.722	0.469	0.9660 (*)	0.873	0.683
V2P_1_2&3	16.8347	0.847	0.829	0.616	0.9870 (*)	0.935	0.8355
FST_1_1&2	0.0484	0.7255	0.7345	0.573	0.3995	0.597	0.0425 (*)
FST_1_2&3	0.0533	0.667	0.737	0.559	0.3885	0.591	0.0475 (*)
LIK_1_1&2	1.2477	0.242	0.338	0.241	0.3765	0.365	0.0480 (*)
DM2_1_1&2	5.0991	0.9620 (*)	0.9570 (*)	0.898	0.9750 (*)	0.9790 (*)	0.913
DM2_1_2&3	3.638	0.904	0.895	0.739	0.9580 (*)	0.942	0.754
NHA_2_3	1	0.0475 (*)	0.064	0.088	0.055	0.057	0.1035
NSS_2_3	0	0.0475 (*)	0.064	0.088	0.055	0.057	0.1035
MPD_2_3	0	0.0475 (*)	0.064	0.088	0.055	0.057	0.1035
VPD_2_3	0	0.0475 (*)	0.064	0.088	0.055	0.057	0.1035
MNS_2_3	0	0.0475 (*)	0.064	0.088	0.055	0.057	0.1035
HST_2_2&3	0.7656	0.9920 (**)	0.9790 (*)	0.9900 (**)	0.9820 (*)	0.9940 (**)	0.9660 (*)

The probability $\text{Prob.}(S_{\text{simul.}} < S_{\text{obs.}})$ given for each summary statistic was calculated from 1,000 virtual datasets simulated from the posterior distributions of parameters obtained under the focused scenario. Corresponding tail-area probabilities (*p*-values) were obtained as $\text{Prob.}(S_{\text{simul.}} < S_{\text{obs.}})$ and $1.0 - \text{Prob.}(S_{\text{simul.}} < S_{\text{obs.}})$ for $\text{Prob.}(S_{\text{simul.}} < S_{\text{obs.}}) \leq 0.5$ and > 0.5 , respectively (*, **, *** = tail-area probability < 0.05 , < 0.01 and < 0.001 , respectively). In addition to the statistics used during the model choice procedure, the model check procedure used also two samples statistics including the mean number of alleles, mean genetic diversity, and mean size variance for microsatellite loci; and for mtDNA the within sample statistics including mean number of the rarest nucleotide at segregating sites and its variance, and the two samples statistics comprising the number of haplotypes and number of segregating sites. Only significant summary statistics for at least one scenario are shown. Abbreviations for the summary statistics are as follows: Mean size variance (VAR); Mean Garcia-Williamson index (MGW); Variance of pairwise difference (V2P); F_{ST} between 2 samples (FST); $d\mu^2$ distance (DM2); Classification index (LIK); Number of mtDNA haplotypes (NHA); Number of segregating sites (NSS); Mean pairwise difference (MPD); Variance of pairwise difference (VPD); Mean number of the rarest nucleotide at segregating sites (MSN); Hudson's FST-statistics (HST).

Table S9. Model selection procedure and performance analysis for the ABC step 2*bii* (figure 2*bii*).

	SC1	SC2
Post. Pr	21.8%	78.2%
95%CI	[21.1 – 22.5]	[77.5 – 78.9]
Performance analysis (based on 1000 simulations)		
Decision	SC1	SC2
D1	86.50%	14.50%
D2	13.50%	85.50%
Model check (number of outlying statistics)		
$P < 0.05$	0	0
$P < 0.01$	1	1
$P < 0.001$	0	0

(see legend of table S5)

Table S10. Model checking for the analysis for the ABC step 2*bii* (figure 2*bii*).

	Obs. value	SC1	SC2
HST_2_2&3	0.7656	0.9950(**)	0.9950(**)

(See legend table S4)

Table S11. Parameter estimation for scenario SC2 in natural units (Fig. 2b_{ii}).

	mean	mode	Q_{2.5}	Q_{5.0}	Q_{25.0}	median	Q_{75.0}	Q_{95.0}	Q_{97.5}
N_{14-PY}	49.6	35.3	11.3	14.6	29.7	46.5	68.3	92.7	96.0
N_{24-TL}	54.8	32.3	11.4	15.5	32.9	53.8	77.2	95.8	97.9
$N_{34-XCSS}$	42.3	13.9	6.9	8.8	20.0	36.3	62.5	91.0	95.4
N_{iso14}	2,520.0	2,090.0	599.0	772.0	1,580.0	2,380.0	3,410.0	4,620.0	4,810.0
N_{iso24}	2,450.0	1,720.0	591.0	756.0	1,530.0	2,300.0	3,300.0	4,600.0	4,790.0
N_{iso34}	2,420.0	2,030.0	578.0	758.0	1,470.0	2,250.0	3,280.0	4,600.0	4,800.0
N_{exp4}	6,580.0	5,660.0	2,900.0	3,360.0	5,090.0	6,570.0	8,170.0	9,620.0	9,840.0
N_{bot4}	6.8	10.0	1.8	2.6	5.1	7.1	8.7	10.0	10.0
N_{anc4}	13,300.0	18,700.0	3,480.0	4,830.0	10,100.0	13,900.0	17,200.0	19,500.0	19,700.0
$T_{crash14}$	3.0	1.0	1.0	1.0	1.5	2.9	4.5	5.0	5.0
$T_{crash24}$	2.6	1.0	1.0	1.0	1.2	2.2	3.7	5.0	5.0
$T_{crash34}$	3.2	5.0	1.0	1.0	2.1	3.4	4.6	5.0	5.0
T_{isol4}	154.0	103.0	21.4	31.4	76.7	126.0	200.0	365.0	448.0
T_{exp4}	1,060.0	340.0	130.0	178.0	394.0	682.0	1,320.0	3,410.0	4,100.0
$db4$	6.1	6.4	1.1	1.9	4.1	6.2	8.1	10.0	10.0
μ_{mic_1}	5.81E-04	4.48E-04	2.13E-04	2.48E-04	4.11E-04	5.68E-04	7.48E-04	9.43E-04	9.76E-04
p_{mic_1}	2.48E-01	3.00E-01	1.26E-01	1.43E-01	2.19E-01	2.62E-01	2.89E-01	3.00E-01	3.00E-01
sni_{mic_1}	1.05E-06	1.00E-08	1.13E-08	1.31E-08	4.49E-08	2.16E-07	1.07E-06	5.54E-06	7.33E-06
u_{seq_2}	9.96E-07	5.61E-07	2.33E-07	2.94E-07	5.41E-07	8.16E-07	1.23E-06	2.27E-06	2.86E-06
$k1_{seq_2}$	1.02E+01	1.69E+01	5.52E-01	1.03E+00	5.10E+00	1.03E+01	1.52E+01	1.90E+01	1.95E+01

Evolutionary scenario SC2 are represented in Fig. 2bii.; Effective population size is provided in term of diploid individuals;
Time of the event are provided in generation before present. Q_x: x % quantile.

Table S12. Composite parameter for the scenario SC2 (see Fig. 2*bii*)

Parameter	mean	mode	Q _{2.5}	Q _{5.0}	Q _{25.0}	median	Q _{75.0}	Q _{95.0}	Q _{97.5}
N14(u+sni)_1	2.48E-02	1.36E-02	3.96E-03	5.41E-03	1.23E-02	2.05E-02	3.28E-02	5.95E-02	6.98E-02
N24(u+sni)_1	3.12E-02	1.16E-03	9.00E-04	1.46E-03	7.78E-03	2.23E-02	4.91E-02	8.86E-02	9.51E-02
N34(u+sni)_1	2.49E-02	1.09E-02	3.22E-03	4.56E-03	1.11E-02	1.99E-02	3.40E-02	6.34E-02	7.30E-02
Niso14(u+sni)_1	2.87E+00	4.98E+00	2.99E-01	4.84E-01	1.71E+00	3.00E+00	4.09E+00	4.94E+00	4.98E+00
Niso24(u+sni)_1	1.49E+00	1.05E+00	3.27E-01	4.27E-01	8.34E-01	1.26E+00	1.93E+00	3.43E+00	3.90E+00
Niso34(u+sni)_1	1.15E+00	3.82E-01	1.22E-01	1.69E-01	4.39E-01	8.26E-01	1.54E+00	3.30E+00	3.96E+00
Nexp4(u+sni)_1	4.88E+00	3.37E+00	3.82E-01	6.40E-01	2.55E+00	4.71E+00	7.20E+00	9.58E+00	9.98E+00
Nbot4(u+sni)_1	3.73E-03	2.99E-03	9.11E-04	1.12E-03	2.32E-03	3.47E-03	4.91E-03	7.23E-03	7.96E-03
Nanc4(u+sni)_1	4.32E+00	9.95E-01	2.54E-01	3.96E-01	1.41E+00	3.08E+00	5.99E+00	1.27E+01	1.51E+01
Tcrash14(u+sni)_1	1.28E-03	6.16E-04	2.19E-04	2.68E-04	5.79E-04	1.01E-03	1.72E-03	3.29E-03	3.82E-03
Tcrash24(u+sni)_1	2.14E-03	1.07E-04	1.38E-04	2.08E-04	9.14E-04	1.96E-03	3.23E-03	4.68E-03	4.94E-03
Tcrash34(u+sni)_1	1.29E-03	1.09E-04	1.14E-04	1.25E-04	2.53E-04	6.67E-04	1.88E-03	4.50E-03	4.93E-03
Tisol4(u+sni)_1	8.39E-02	3.21E-02	1.11E-02	1.49E-02	3.61E-02	6.20E-02	1.05E-01	2.18E-01	2.84E-01
Texp4(u+sni)_1	1.92E-01	3.05E-02	1.99E-02	2.28E-02	4.45E-02	8.66E-02	1.98E-01	6.94E-01	1.04E+00
db4(u+sni)_1	3.87E-03	1.49E-03	5.99E-04	8.27E-04	2.03E-03	3.51E-03	5.43E-03	7.99E-03	8.63E-03
N14useq_2	4.92E-05	2.32E-05	5.63E-06	8.53E-06	2.15E-05	3.72E-05	6.18E-05	1.28E-04	1.66E-04
N24useq_2	5.58E-05	2.39E-06	1.24E-06	1.95E-06	7.34E-06	2.01E-05	5.53E-05	2.42E-04	3.69E-04
N34useq_2	2.49E-05	7.08E-07	5.47E-07	7.63E-07	2.99E-06	8.41E-06	2.33E-05	1.00E-04	1.64E-04
Niso14useq_2	2.01E-03	9.26E-04	3.75E-04	4.82E-04	9.97E-04	1.57E-03	2.45E-03	4.92E-03	6.25E-03
Niso24useq_2	2.42E-03	1.06E-03	3.29E-04	4.38E-04	9.91E-04	1.69E-03	2.85E-03	6.69E-03	9.39E-03
Niso34useq_2	3.96E-03	1.97E-04	8.98E-05	1.45E-04	6.19E-04	1.67E-03	4.47E-03	1.68E-02	2.34E-02
Nexp4useq_2	9.61E-03	9.26E-04	4.98E-04	6.61E-04	1.84E-03	4.27E-03	1.06E-02	4.01E-02	5.57E-02
Nbot4useq_2	4.77E-05	1.98E-07	1.24E-06	2.53E-06	1.79E-05	4.64E-05	7.65E-05	9.80E-05	9.83E-05
Nanc4useq_2	3.52E-02	1.67E-02	2.84E-03	4.34E-03	1.36E-02	2.65E-02	4.82E-02	9.75E-02	1.16E-01
Tcrash14useq_2	5.67E-06	1.49E-06	6.54E-07	9.02E-07	2.31E-06	4.27E-06	7.48E-06	1.52E-05	1.90E-05
Tcrash24useq_2	2.63E-06	1.28E-06	3.89E-07	5.05E-07	1.15E-06	1.98E-06	3.30E-06	6.80E-06	8.71E-06
Tcrash34useq_2	3.33E-06	1.06E-06	4.65E-07	6.09E-07	1.43E-06	2.54E-06	4.28E-06	8.69E-06	1.08E-05
Tisol4useq_2	1.28E-04	4.49E-05	1.47E-05	2.12E-05	5.28E-05	9.44E-05	1.63E-04	3.43E-04	4.32E-04
Texp4useq_2	3.15E-03	6.41E-05	5.07E-05	8.23E-05	3.87E-04	1.14E-03	3.29E-03	1.37E-02	1.99E-02
db4useq_2	7.99E-06	2.42E-07	2.02E-07	2.93E-07	1.27E-06	3.57E-06	9.78E-06	3.10E-05	4.29E-05

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