

Example: Microsatellite data set

MIGRATION RATE AND POPULATION SIZE ESTIMATION

using the coalescent and maximum likelihood or Bayesian inference

Migrate-n version 3.3.2 [2091]

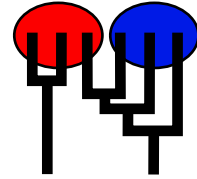
Compiled for a PARALLEL COMPUTER ARCHITECTURE

One master and 2 compute nodes are available.

Compiled for a SYMMETRIC MULTIPROCESSORS

Program started at Tue Nov 20 16:45:50 2012

Program finished at Tue Nov 20 16:46:38 2012



Options

Datatype: Microsatellite data [Brownian motion]
Missing data: not included

Inheritance scalers in use for Thetas:

All loci use an inheritance scaler of 1.0

[The locus with a scaler of 1.0 used as reference]

Random number seed: (from parmfile) 310705631

Start parameters:

Theta values were generated RANDOM start value from U(min,msx)

M values were generated from the FST-calculation

Connection type matrix:

where m = average (average over a group of Thetas or M,

s = symmetric M, S = symmetric 4Nm, 0 = zero, and not estimated,

* = free to vary, Thetas are on diagonal

Population	1	2
1 population_num	*	0
2 population_num	*	*

Order of parameters:

1	Θ_1	<displayed>
2	Θ_2	<displayed>

4 M $1 \rightarrow 2$ <displayed>

Mutation rate among loci:

Mutation rate is constant for all loci

Analysis strategy:

Bayesian inference

Proposal distributions for parameter

Parameter	Proposal
Theta	Slice sampling
M	Slice sampling

Prior distribution for parameter

Parameter	Prior	Minimum	Mean*	Maximum	Delta	Bins
Theta	Uniform	0.000000	10.000000	20.000000	2.000000	500
M	Uniform	0.000000	10.000000	20.000000	2.000000	500

Markov chain settings:

Long chain

Number of chains	1
Recorded steps [a]	5000
Increment (record every x step [b])	1
Number of concurrent chains (replicates) [c]	2
Visited (sampled) parameter values [a*b*c]	10000
Number of discard trees per chain (burn-in)	10000

Multiple Markov chains:

Static heating scheme

4 chains with temperatures
 1000000.00 3.00 1.50 1.00
 Swapping interval is 1

Print options:

Data file:	infile.msat
Output file:	outfile-bayes
Posterior distribution raw histogram file:	bayesfile
Print data:	No
Print genealogies [only some for some data type]:	None

Data summary

Datatype: Microsatellite data
 [Data was used as repeat-length information]
 Number of loci: 10

Population	Locus	Gene copies data	(missing)
1 population_number___0	1	50	(0)
	2	50	(0)
	3	50	(0)
	4	50	(0)
	5	50	(0)
	6	50	(0)
	7	50	(0)
	8	50	(0)
	9	50	(0)
	10	50	(0)
2 population_number___1	1	42	(0)
	2	42	(0)
	3	42	(0)
	4	42	(0)
	5	42	(0)
	6	42	(0)
	7	42	(0)
	8	42	(0)
	9	42	(0)
	10	42	(0)
Total of all populations	1	92	(0)
	2	92	(0)
	3	92	(0)
	4	92	(0)
	5	92	(0)
	6	92	(0)
	7	92	(0)
	8	92	(0)
	9	92	(0)
	10	92	(0)

Allele frequency spectra

Locus 1

Allele	Pop1	Pop2	All
16	0.220	0.167	0.196
19	0.040	0.071	0.054
18	0.060	0.119	0.087
15	0.220	0.024	0.130
21	0.020	0.167	0.087
23	0.020	0.119	0.065
17	0.280	0.095	0.196
22	0.060	0.119	0.087
25	0.060	0.024	0.043
24	0.020	-	0.011
26	-	0.024	0.011
27	-	0.048	0.022
29	-	0.024	0.011
Total	10	12	13
H _{exp}	0.811	0.883	0.874

Locus 2

Allele	Pop1	Pop2	All
16	0.520	0.571	0.543
19	0.040	-	0.022
18	0.220	0.119	0.174
17	0.160	0.167	0.163
15	0.020	-	0.011
21	0.020	0.071	0.043
20	0.020	0.024	0.022
22	-	0.048	0.022
Total	7	6	8
H _{exp}	0.653	0.624	0.518

Locus 3

Allele	Pop1	Pop2	All
19	0.240	0.262	0.250
20	0.280	0.476	0.370
18	0.080	0.095	0.087
21	0.280	0.119	0.207

Allele	Pop1	Pop2	All
22	0.120	0.048	0.087
Total	5	5	5
H _{exp}	0.765	0.679	0.261
Locus 4			
Allele	Pop1	Pop2	All
16	0.080	0.071	0.076
24	0.180	0.024	0.109
15	0.020	0.048	0.033
25	0.160	0.167	0.163
14	0.020	0.048	0.033
19	0.100	0.143	0.120
12	0.060	-	0.033
20	0.080	0.190	0.130
23	0.060	0.119	0.087
28	0.020	-	0.011
22	0.060	0.024	0.043
21	0.160	0.119	0.141
13	-	0.024	0.011
26	-	0.024	0.011
Total	12	12	14
H _{exp}	0.882	0.875	0.153
Locus 5			
Allele	Pop1	Pop2	All
20	0.400	0.524	0.457
21	0.420	0.357	0.391
19	0.180	0.119	0.152
Total	3	3	3
H _{exp}	0.631	0.584	-0.232
Locus 6			
Allele	Pop1	Pop2	All
19	0.060	-	0.033
20	0.100	0.024	0.065
18	0.300	0.214	0.261
22	0.200	0.119	0.163
21	0.120	0.476	0.283
16	0.060	-	0.033
24	0.160	0.048	0.109

Allele	Pop1	Pop2	All
17	-	0.119	0.054
Total	7	6	8
H _{exp}	0.813	0.696	-0.427
Locus 7			
Allele	Pop1	Pop2	All
23	0.040	0.238	0.130
20	0.660	0.143	0.424
22	0.180	0.190	0.185
21	0.100	0.333	0.207
19	0.020	0.095	0.054
Total	5	5	5
H _{exp}	0.520	0.766	-0.704
Locus 8			
Allele	Pop1	Pop2	All
19	0.520	0.524	0.522
17	0.040	0.048	0.043
18	0.100	0.071	0.087
20	0.140	0.190	0.163
16	0.080	-	0.043
22	0.100	0.048	0.076
15	0.020	0.048	0.033
23	-	0.071	0.033
Total	7	7	8
H _{exp}	0.682	0.672	-1.022
Locus 9			
Allele	Pop1	Pop2	All
24	0.080	0.024	0.054
19	0.300	0.429	0.359
20	0.300	0.167	0.239
23	0.180	0.143	0.163
22	0.080	0.024	0.054
18	0.020	0.071	0.043
21	0.040	0.095	0.065
25	-	0.048	0.022
Total	7	8	8
H _{exp}	0.773	0.751	-1.247

Locus 10

Allele	Pop1	Pop2	All
22	0.100	0.214	0.152
20	0.440	0.214	0.337
23	0.080	0.167	0.120
24	0.020	-	0.011
19	0.160	0.167	0.163
21	0.060	0.048	0.054
18	0.080	-	0.043
15	0.020	0.071	0.043
17	0.040	0.048	0.043
25	-	0.071	0.033
Total	9	8	10
H_{exp}	0.752	0.838	-1.434

Average expected heterozygosity

	Pop1	Pop2	All
H_{exp}	1.456	1.474	0.016

Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	Θ_1	8.48000	13.08000	0.00000	13.76000	0.00000	0.00000	13.67752
1	Θ_2	8.36000	15.92000	0.00000	16.92000	0.00000	14.46000	14.31475
1	$M_{1 \rightarrow 2}$	0.000	0.240	0.000	0.720	0.000	0.620	0.631
2	Θ_1	3.68000	5.04000	0.00000	8.28000	0.00000	0.00000	10.76360
2	Θ_2	1.24000	1.56000	0.00000	3.32000	0.00000	3.58000	4.34140
2	$M_{1 \rightarrow 2}$	0.000	0.320	0.000	1.520	0.000	2.260	2.331
3	Θ_1	3.36000	4.52000	0.00000	5.76000	0.00000	0.00000	6.62961
3	Θ_2	1.20000	1.36000	0.00000	4.72000	0.00000	4.58000	7.45518
3	$M_{1 \rightarrow 2}$	0.000	1.760	0.000	3.800	0.000	3.500	4.121
4	Θ_1	10.84000	19.36000	0.00000	20.00000	0.00000	0.00000	10.43002
4	Θ_2	2.16000	2.44000	0.00000	5.24000	0.00000	6.10000	9.10683
4	$M_{1 \rightarrow 2}$	0.000	1.560	0.000	2.560	0.000	2.140	2.100
5	Θ_1	0.92000	1.80000	0.00000	2.88000	0.00000	0.00000	2.41524
5	Θ_2	0.36000	0.84000	0.00000	2.08000	0.00000	1.86000	2.25297
5	$M_{1 \rightarrow 2}$	0.000	6.520	0.000	8.920	0.000	8.780	9.183
6	Θ_1	2.12000	2.48000	0.00000	4.68000	0.00000	0.00000	5.18615
6	Θ_2	0.00000	0.24000	0.00000	0.84000	0.00000	0.70000	0.69108
6	$M_{1 \rightarrow 2}$	0.000	19.240	0.000	19.960	0.000	14.380	14.148
7	Θ_1	2.16000	3.44000	0.00000	4.80000	0.00000	0.00000	4.06002
7	Θ_2	2.52000	2.84000	0.00000	7.04000	0.00000	9.42000	10.43260
7	$M_{1 \rightarrow 2}$	0.000	1.280	0.000	2.600	0.000	2.260	2.322
8	Θ_1	4.48000	6.36000	0.00000	9.24000	0.00000	0.00000	8.57411
8	Θ_2	1.72000	1.92000	0.00000	5.00000	0.00000	8.06000	8.74874
8	$M_{1 \rightarrow 2}$	0.000	2.040	0.000	4.360	0.000	3.900	4.302
9	Θ_1	6.96000	11.20000	0.00000	13.08000	0.00000	0.00000	12.09407
9	Θ_2	3.80000	5.92000	0.00000	10.52000	0.00000	9.34000	10.02862
9	$M_{1 \rightarrow 2}$	0.000	1.560	0.000	4.840	0.000	5.620	7.064

10	Θ_1	4.92000	6.04000	0.00000	8.68000	0.00000	0.00000	12.17507
10	Θ_2	7.20000	9.20000	0.00000	13.12000	0.00000	12.58000	12.81515
10	$M_{1 \rightarrow 2}$	0.000	0.200	0.000	0.640	0.000	0.500	0.486
<hr/>								
All	Θ_1	0.68000	0.80000	1.10000	1.32000	1.44000	1.26000	1.65862
All	Θ_2	0.00000	0.00000	0.06000	0.36000	0.92000	0.38000	0.28477
All	$M_{1 \rightarrow 2}$	2.400	2.680	2.900	3.040	3.320	2.660	1.789

Citation suggestions:

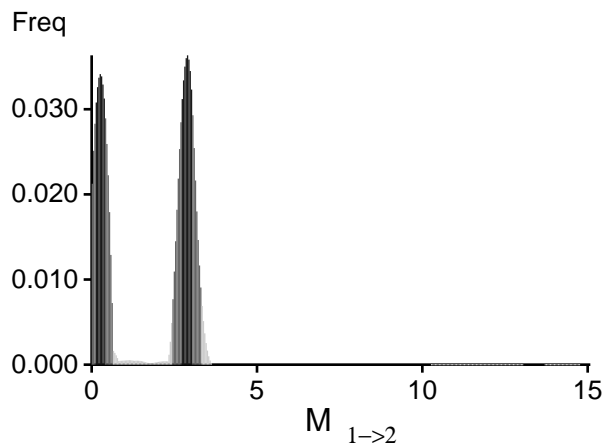
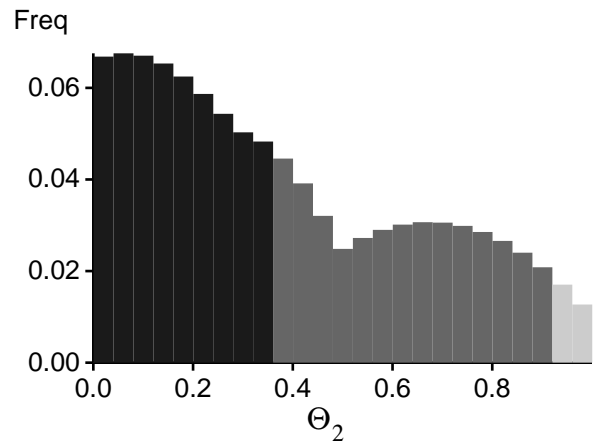
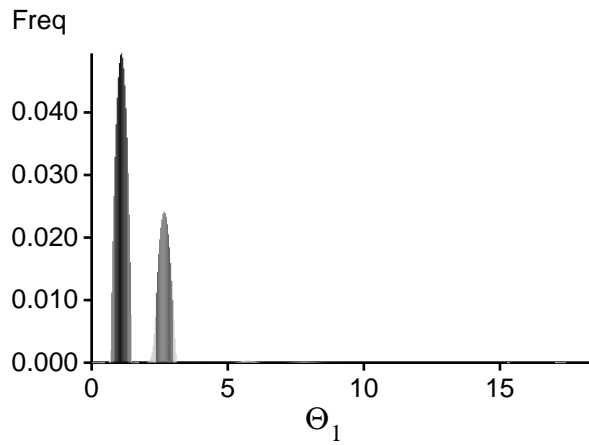
Beerli P., 2006. Comparison of Bayesian and maximum-likelihood inference of population genetic parameters.

Bioinformatics 22:341-345

Beerli P., 2009. How to use MIGRATE or why are Markov chain Monte Carlo programs difficult to use?

In Population Genetics for Animal Conservation, G. Bertorelle, M. W. Bruford, H. C. Hauffe, A. Rizzoli, and C. Vernesi, eds., vol. 17 of Conservation Biology, Cambridge University Press, Cambridge UK, pp. 42-79.

Bayesian Analysis: Posterior distribution over all loci



Log-Probability of the data given the model (marginal likelihood)

Use this value for Bayes factor calculations:

$BF = \text{Exp}[\ln(\text{Prob}(D \mid \text{thisModel}) - \ln(\text{Prob}(D \mid \text{otherModel}))]$

or as $LBF = 2 (\ln(\text{Prob}(D \mid \text{thisModel}) - \ln(\text{Prob}(D \mid \text{otherModel})))$

shows the support for thisModel]

Locus	Raw thermodynamic score(1a)	Bezier approximation score(1b)	Harmonic mean(2)
1	-2521.69	-533.71	-115.99
2	-1904.81	-401.07	-87.67
3	-408.49	-165.33	-73.98
4	-2544.88	-527.68	-101.96
5	-215.65	-99.86	-36.48
6	-4684.51	-831.50	-50.61
7	-880.12	-230.81	-63.80
8	-967.52	-257.16	-73.18
9	-3363.37	-647.07	-93.68
10	-2021.63	-443.62	-105.64
All	-19576.29	-4201.40	-866.59

(1a, 1b and 2) are approximations to the marginal likelihood, make sure that the program run long enough!

(1a, 1b) and (2) should give similar results, in principle.

But (2) is overestimating the likelihood, it is presented for historical reasons and should not be used

(1a, 1b) needs heating with chains that span a temperature range of 1.0 to at least 100,000.

(1b) is using a Bezier-curve to get better approximations for runs with low number of heated chains

[Scaling factor = -63.599643

Citation suggestions:

Beerli P. and M. Palczewski, 2010. Unified framework to evaluate panmixia and migration direction among multiple sampling locations, *Genetics*, 185: 313-326.

Acceptance ratios for all parameters and the genealogies

Parameter	Accepted changes	Ratio
Θ_1	248770/248770	1.00000
Θ_2	252330/252330	1.00000
$M_{1 \rightarrow 2}$	248825/248825	1.00000
Genealogies	80570/249775	0.32257

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sample Size
Θ_1	0.94358	14235.43
Θ_2	0.91181	20497.37
$M_{1 \rightarrow 2}$	0.92998	21932.81
$\text{Ln}[\text{Prob}(D G)]$	0.99394	1680.90

Potential Problems

This section reports potential problems with your run, but such reporting is often not very accurate. With many parameters in a multilocus analysis, it is very common that some parameters for some loci will not be very informative, triggering suggestions (for example to increase the prior range) that are not sensible. This suggestion tool will improve with time, therefore do not blindly follow its suggestions. If some parameters are flagged, inspect the tables carefully and judge whether an action is required. For example, if you run a Bayesian inference with sequence data, for macroscopic species there is rarely the need to increase the prior for Theta beyond 0.1; but if you use microsatellites it is rather common that your prior distribution for Theta should have a range from 0.0 to 100 or more. With many populations (>3) it is also very common that some migration routes are estimated poorly because the data contains little or no information for that route. Increasing the range will not help in such situations, reducing number of parameters may help in such situations.

No warning was recorded during the run