## **AUTO**

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

Bayesian inference using the structured coalescent

Migrate-n version 5.0.0a [May-20-2017]

Using Intel AVX (Advanced Vector Extensions)

Compiled for PARALLEL computer architectures

One master and 100 compute nodes are available.

Program started at Sun Aug 13 13:24:17 2017

Program finished at Sun Aug 13 14:28:50 2017 [Runtime:0000:01:04:33]



### **Options**

Datatype: DNA sequence data

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: (with internal timer) 1831315036

Start parameters:

Theta values were generated Using a percent value of the prior

M values were generated Using a percent value of the prior

Connection matrix:

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

s = symmetric migration M, S = symmetric 4Nm,

0 = zero, and not estimated,

\* = migration free to vary, Thetas are on diagonal

1

d = row population split off column population, D = split and then migration

Population

1 Romanshorn 0 \*

Order of parameters:

1  $\Theta_1$  <displayed>

Mutation rate among loci: Mutation rate is constant for all loci

Analysis strategy: Bayesian inference

-Population size estimation: Exponential Distribution

Proposal distributions for parameter

Parameter Proposal
Theta Metropolis sampling
M Metropolis sampling
Divergence Metropolis sampling
Divergence Spread Metropolis sampling
Genealogy Metropolis-Hastings

Prior distribution for parameter

Parameter Prior Minimum MeanMaximum Delta Bins UpdateFreq
1 Theta -11 Uniform 0.000000 0.050 0.100 0.010 1500 0.20000

[-1 -1 means priors were set globally]

Markov chain settings:

Long chain

Number of chains1Recorded steps [a]50000Increment (record every x step [b]200Number of concurrent chains (replicates) [c]2

Visited (sampled) parameter values [a\*b\*c] 20000000

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.0.4 Haplotyping is turned on: NO

Output file: outfile\_0.4\_0.7

Posterior distribution raw histogram file: bayesfile
Raw data from the MCMC run: bayesallfile\_0.4\_0.7

Print data:

Print genealogies [only some for some data type]:

### Data summary

Data file: infile.0.4
Datatype: Sequence data
Number of loci: 100

Mutatior	nmodel:			
Locus S	Sublocus	Mutationmodel	Mutationmodel parameters	
	4	lulus Cantan	[December 0.05]	
1	1	Jukes-Cantor	[Basefreq: =0.25]	
2	1	Jukes-Cantor	[Basefreq: =0.25]	
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Jukes-Cantor

Jukes-Cantor

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Locus S	Sublocus Region type	Rate of change	Probability	Patch size	
1	1 1	1.000	1.000	1.000	
2	1 1	1.000	1.000	1.000	
3	1 1	1.000	1.000	1.000	
4	1 1	1.000	1.000	1.000	
5	1 1	1.000	1.000	1.000	
6	1 1	1.000	1.000	1.000	

7	1	1	1.000	1.000	1.000	
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9	1	1	1.000	1.000	1.000	
10	1	1	1.000	1.000	1.000	
11	1	1	1.000	1.000	1.000	
12	1	1	1.000	1.000	1.000	
13	1	1	1.000	1.000	1.000	
14	1	1	1.000	1.000	1.000	
15	1	1	1.000	1.000	1.000	
16	1	1	1.000	1.000	1.000	
17	1	1	1.000	1.000	1.000	
18	1	1	1.000	1.000	1.000	
19	1	1	1.000	1.000	1.000	
20	1	1	1.000	1.000	1.000	
21	1	1	1.000	1.000	1.000	
22	1	1	1.000	1.000	1.000	
23	1	1	1.000	1.000	1.000	
24	1	1	1.000	1.000	1.000	
25	1	1	1.000	1.000	1.000	
26	1	1	1.000	1.000	1.000	
27	1	1	1.000	1.000	1.000	
28	1	1	1.000	1.000	1.000	
29	1	1	1.000	1.000	1.000	
30	1	1	1.000	1.000	1.000	
31	1	1	1.000	1.000	1.000	
32	1	1	1.000	1.000	1.000	
33	1	1	1.000	1.000	1.000	
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35	1	1	1.000	1.000	1.000	
36	1	1	1.000	1.000	1.000	
37	1	1	1.000	1.000	1.000	
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39	1	1	1.000	1.000	1.000	
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42	1	1	1.000	1.000	1.000	
43	1	1	1.000	1.000	1.000	
44	1	1	1.000	1.000	1.000	
45	1	1	1.000	1.000	1.000	
46	1	1	1.000	1.000	1.000	
47	1	1	1.000	1.000	1.000	
48	1	1	1.000	1.000	1.000	
49	1	1	1.000	1.000	1.000	
50	1	1	1.000	1.000	1.000	
51	1	1	1.000	1.000	1.000	

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52	1	1	1.000	1.000	1.000	
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56	1	1	1.000	1.000	1.000	
57	1	1	1.000	1.000	1.000	
58	1	1	1.000	1.000	1.000	
59	1	1	1.000	1.000	1.000	
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62	1	1	1.000	1.000	1.000	
63	1	1	1.000	1.000	1.000	
64	1	1	1.000	1.000	1.000	
65	1	1	1.000	1.000	1.000	
66	1	1	1.000	1.000	1.000	
67	1	1	1.000	1.000	1.000	
68	1	1	1.000	1.000	1.000	
69	1	1	1.000	1.000	1.000	
70	1	1	1.000	1.000	1.000	
71	1	1	1.000	1.000	1.000	
72	1	1	1.000	1.000	1.000	
73	1	1	1.000	1.000	1.000	
74	1	1	1.000	1.000	1.000	
75	1	1	1.000	1.000	1.000	
76	1	1	1.000	1.000	1.000	
77	1	1	1.000	1.000	1.000	
78	1	1	1.000	1.000	1.000	
79	1	1	1.000	1.000	1.000	
80	1	1	1.000	1.000	1.000	
81	1	1	1.000	1.000	1.000	
82	1	1	1.000	1.000	1.000	
83	1	1	1.000	1.000	1.000	
84	1	1	1.000	1.000	1.000	
85	1	1	1.000	1.000	1.000	
86	1	1	1.000	1.000	1.000	
87	1	1	1.000	1.000	1.000	
88	1	1	1.000	1.000	1.000	
89	1	1	1.000	1.000	1.000	
90	1	1	1.000	1.000	1.000	
91	1	1	1.000	1.000	1.000	
92	1	1	1.000	1.000	1.000	
93	1	1	1.000	1.000	1.000	
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97	1	1	1.000	1.000	1.000	
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	98	10
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1	100	10

# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00495
2	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00494
3	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
4	$\Theta_1$	0.00127	0.00427	0.00657	0.00960	0.01913	0.00817	0.00915
5	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
6	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
7	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
8	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00494
9	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
10	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00495
11	$\Theta_1$	0.00033	0.00287	0.00470	0.00713	0.01453	0.00597	0.00668
12	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00497
13	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00495
14	$\Theta_1$	0.00033	0.00287	0.00463	0.00693	0.01380	0.00577	0.00642
15	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
16	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00494
17	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00496
18	$\Theta_1$	0.00133	0.00153	0.00410	0.00760	0.00793	0.00497	0.00552

19	$\Theta_1$	0.00007	0.00247	0.00417	0.00627	0.01253	0.00517	0.00577
20	$\Theta_1$	0.00107	0.00280	0.00370	0.00460	0.00720	0.00450	0.00496
21	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00495
22	$\Theta_1$	0.00107	0.00207	0.00370	0.00547	0.00713	0.00450	0.00496
23	$\Theta_1$	0.00247	0.00620	0.00990	0.01533	0.03080	0.01330	0.01552
24	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00495
25	$\Theta_1$	0.00007	0.00247	0.00417	0.00627	0.01247	0.00517	0.00575
26	$\Theta_1$	0.00000	0.00340	0.00577	0.00913	0.03280	0.00783	0.00937
27	$\Theta_1$	0.00020	0.00207	0.00370	0.00547	0.00953	0.00450	0.00495
28	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00497
29	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
30	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00497
31	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00495
32	$\Theta_1$	0.00107	0.00247	0.00363	0.00493	0.00713	0.00450	0.00494
33	$\Theta_1$	0.00107	0.00273	0.00370	0.00473	0.00720	0.00450	0.00495
34	$\Theta_1$	0.00100	0.00280	0.00463	0.00693	0.01107	0.00577	0.00643
35	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
36	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
37	$\Theta_1$	0.00007	0.00240	0.00410	0.00607	0.01207	0.00503	0.00559
38	$\Theta_1$	0.00020	0.00267	0.00443	0.00667	0.01333	0.00550	0.00615
39	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
40	$\Theta_1$	0.00007	0.00240	0.00410	0.00613	0.01213	0.00503	0.00561
41	$\Theta_1$	0.00007	0.00247	0.00417	0.00627	0.01247	0.00517	0.00575

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
43	$\Theta_1$	0.00013	0.00247	0.00417	0.00620	0.01233	0.00510	0.00568
44	$\Theta_1$	0.00033	0.00313	0.00537	0.00873	0.02100	0.00750	0.00902
45	$\Theta_1$	0.00313	0.00587	0.00750	0.00940	0.01527	0.00950	0.01067
46	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
47	$\Theta_1$	0.00087	0.00360	0.00563	0.00827	0.01647	0.00697	0.00778
48	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
49	$\Theta_1$	0.00053	0.00207	0.00370	0.00547	0.00867	0.00450	0.00497
50	$\Theta_1$	0.00107	0.00673	0.00783	0.00920	0.03100	0.01077	0.01279
51	$\Theta_1$	0.00007	0.00240	0.00410	0.00607	0.01200	0.00503	0.00556
52	$\Theta_1$	0.00033	0.00280	0.00457	0.00673	0.01307	0.00557	0.00616
53	$\Theta_1$	0.00007	0.00240	0.00410	0.00607	0.01207	0.00503	0.00560
54	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00494
55	$\Theta_1$	0.00107	0.00207	0.00370	0.00547	0.00713	0.00450	0.00494
56	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
57	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
58	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
59	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00494
60	$\Theta_1$	0.00147	0.00453	0.00470	0.00480	0.01013	0.00590	0.00661
61	$\Theta_1$	0.00087	0.00360	0.00570	0.00847	0.01707	0.00717	0.00803

62	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00495
63	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00495
64	$\Theta_1$	0.00073	0.00347	0.00543	0.00800	0.01587	0.00670	0.00748
65	$\Theta_1$	0.00020	0.00267	0.00443	0.00660	0.01333	0.00550	0.00613
66	$\Theta_1$	0.00007	0.00247	0.00417	0.00627	0.01247	0.00517	0.00576
67	$\Theta_1$	0.00020	0.00260	0.00437	0.00647	0.01287	0.00537	0.00595
68	$\Theta_1$	0.00140	0.00313	0.00417	0.00527	0.00813	0.00517	0.00577
69	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
70	$\Theta_1$	0.00013	0.00247	0.00417	0.00620	0.01233	0.00510	0.00568
71	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01607	0.00450	0.00496
72	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
73	$\Theta_1$	0.00280	0.00580	0.00883	0.01293	0.02253	0.01117	0.01262
74	$\Theta_1$	0.00227	0.00440	0.00683	0.01000	0.01553	0.00850	0.00954
75	$\Theta_1$	0.00193	0.00480	0.00743	0.01113	0.01993	0.00950	0.01069
76	$\Theta_1$	0.00000	0.00213	0.00370	0.00553	0.01073	0.00450	0.00495
77	$\Theta_1$	0.00020	0.00260	0.00443	0.00660	0.01340	0.00550	0.00616
78	$\Theta_1$	0.00020	0.00260	0.00437	0.00647	0.01287	0.00537	0.00596
79	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
80	$\Theta_1$	0.00080	0.00380	0.00630	0.00993	0.02273	0.00850	0.01006
81	$\Theta_1$	0.00020	0.00260	0.00430	0.00647	0.01280	0.00537	0.00593
82	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
83	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00496
84	$\Theta_1$	0.00000	0.00213	0.00370	0.00553	0.01073	0.00450	0.00496

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.00007	0.00240	0.00410	0.00607	0.01200	0.00503	0.00555
86	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01073	0.00450	0.00495
87	$\Theta_1$	0.00020	0.00207	0.00370	0.00547	0.00953	0.00450	0.00494
88	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01067	0.00450	0.00493
89	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00495
90	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
91	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00496
92	$\Theta_1$	0.00253	0.00480	0.00650	0.00853	0.01373	0.00857	0.00986
93	$\Theta_1$	0.00053	0.00313	0.00497	0.00733	0.01453	0.00617	0.00682
94	$\Theta_1$	0.00053	0.00313	0.00503	0.00747	0.01487	0.00623	0.00695
95	$\Theta_1$	0.00020	0.00260	0.00437	0.00660	0.01327	0.00550	0.00612
96	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01087	0.00450	0.00496
97	$\Theta_1$	0.00000	0.00207	0.00370	0.00547	0.01080	0.00450	0.00497
98	$\Theta_1$	0.00273	0.00540	0.00837	0.01227	0.02033	0.01057	0.01196
99	$\Theta_1$	0.00020	0.00253	0.00423	0.00627	0.01253	0.00523	0.00579
100	$\Theta_1$	0.00053	0.00313	0.00503	0.00740	0.01453	0.00617	0.00685
All	$\Theta_1$	0.00160	0.00280	0.00377	0.00467	0.00587	0.00383	0.00376

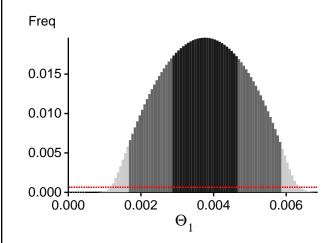
Citation suggestions:

Beerli P., 2006. Comparison of Bayesian and maximum-likelihood inference of population genetic parameters. Bioinformatics 22:341-345

Beerli P., 2007. Estimation of the population scaled mutation rate from microsatellite data, Genetics, 177:1967-1968.

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:  $BF = Exp[\ ln(Prob(D \mid thisModel) - ln(\ Prob(\ D \mid otherModel)) \\ or \ as \ LBF = 2 \ (ln(Prob(D \mid thisModel) - ln(\ Prob(\ D \mid otherModel))) \\ shows the \ support for \ thisModel]$ 

Locus	TI(1a)	BTI(1b)	SS(2)	HS(3)
1	-13908.82	-13736.36	-13772.87	-13866.23
2	-13910.24	-13738.05	-13775.07	-13868.17
3	-13910.79	-13738.18	-13774.68	-13868.56
4	-14011.93	-13835.71	-13883.06	-13964.79
5	-13910.82	-13738.33	-13774.67	-13868.68
6	-13907.89	-13735.41	-13771.94	-13865.08
7	-13906.79	-13734.62	-13768.85	-13864.78
8	-13908.56	-13735.69	-13772.51	-13865.53
9	-13909.42	-13737.02	-13773.57	-13866.89
10	-13907.51	-13735.57	-13772.26	-13865.77
11	-13932.17	-13759.93	-13802.49	-13890.27
12	-13909.63	-13737.08	-13773.33	-13866.71
13	-13910.41	-13738.11	-13773.73	-13868.26
14	-13938.61	-13763.76	-13802.57	-13892.60
15	-13910.11	-13737.88	-13773.83	-13868.01
16	-13909.85	-13737.41	-13774.27	-13868.70
17	-13911.10	-13738.38	-13775.07	-13868.37
18	-13926.03	-13751.69	-13788.43	-13881.05
19	-13922.77	-13750.60	-13790.03	-13881.02
20	-13909.85	-13737.74	-13774.19	-13870.80
21	-13905.26	-13733.10	-13769.95	-13864.70
22	-13909.03	-13736.31	-13772.83	-13866.43
23	-14536.78	-14259.90	-14300.39	-14375.96
24	-13908.97	-13736.48	-13773.29	-13866.76
25	-13919.79	-13747.54	-13786.46	-13878.19
26	-37371.54	-24634.61	-22353.71	-22454.00
27	-13910.14	-13737.70	-13773.90	-13868.10
28	-13907.74	-13735.27	-13772.18	-13865.67
29	-13910.86	-13738.34	-13774.91	-13868.42

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 13:24:17]

30	-13909.37	-13736.70	-13773.15	-13866.71
31	-13910.05	-13737.24	-13774.01	-13866.83
32	-13910.33	-13737.63	-13773.41	-13867.45
33	-13910.23	-13737.92	-13774.53	-13868.04
34	-13937.05	-13762.68	-13802.64	-13892.99
35	-13910.19	-13738.06	-13774.95	-13868.02
36	-13910.59	-13738.29	-13774.98	-13868.97
37	-13925.07	-13751.10	-13787.93	-13880.06
38	-13932.07	-13759.84	-13799.29	-13890.07
39	-13910.26	-13737.93	-13774.46	-13867.94
40	-13924.01	-13750.37	-13787.47	-13880.89
41	-13923.19	-13750.70	-13789.35	-13882.60
42	-13909.39	-13737.34	-13773.98	-13868.18
43	-13924.41	-13751.23	-13789.20	-13881.88
44	-35570.01	-26309.61	-24469.28	-24893.97
45	-13993.96	-13815.61	-13862.85	-13944.33
46	-13910.94	-13738.39	-13774.44	-13868.20
47	-13999.98	-13816.90	-13860.84	-13945.65
48	-13906.95	-13735.05	-13771.76	-13864.83
49	-13910.34	-13738.21	-13775.05	-13868.22
50	-15907.85	-15334.40	-15329.20	-15406.25
51	-13924.71	-13749.05	-13786.07	-13879.06
52	-13963.53	-13778.20	-13817.44	-13906.24
53	-13925.36	-13751.59	-13788.86	-13880.58
54	-13910.02	-13737.24	-13773.72	-13868.49
55	-13906.50	-13734.03	-13770.03	-13864.02
56	-13909.91	-13737.32	-13773.97	-13868.24
57	-13908.69	-13736.15	-13773.11	-13865.99
58	-13906.22	-13734.05	-13770.51	-13865.52
59	-13910.87	-13738.42	-13775.01	-13868.42
60	-13933.26	-13760.53	-13801.29	-13889.94
61	-13963.49	-13785.45	-13827.65	-13914.19
62	-13911.25	-13738.54	-13775.11	-13868.30
63	-13908.34	-13735.92	-13772.12	-13865.93
64	-14060.91	-13853.09	-13892.85	-13979.25
65	-13931.47	-13758.87	-13797.32	-13889.03
66	-13922.36	-13750.34	-13789.26	-13881.37
67	-13934.81	-13759.74	-13798.62	-13890.07
68	-13921.28	-13749.05	-13788.23	-13881.07
69	-13908.53	-13736.13	-13772.56	-13866.44
70	-13924.49	-13751.56	-13788.86	-13881.15
71	-13907.78	-13735.75	-13772.12	-13865.95
72	-13907.43	-13735.00	-13771.60	-13864.79
73	-14020.43	-13843.08	-13893.75	-13973.50
74	-14145.57	-13935.08	-13979.42	-14061.98

75	-14396.61	-14132.86	-14170.87	-14250.09
76	-13909.36	-13737.04	-13771.70	-13867.04
77	-13932.16	-13760.29	-13799.83	-13891.31
78	-13934.33	-13760.00	-13799.04	-13890.15
79	-13909.39	-13737.27	-13773.68	-13868.32
80	-17441.27	-15999.63	-15832.54	-15911.46
81	-13933.55	-13759.02	-13798.09	-13888.50
82	-13910.71	-13738.14	-13774.98	-13868.15
83	-13911.47	-13738.32	-13774.66	-13867.81
84	-13910.16	-13737.95	-13773.78	-13869.34
85	-13925.66	-13750.98	-13786.41	-13880.19
86	-13910.29	-13738.12	-13774.77	-13868.45
87	-13910.57	-13738.23	-13774.28	-13868.13
88	-13908.63	-13736.35	-13773.01	-13866.13
89	-13906.75	-13734.62	-13771.62	-13864.36
90	-13910.29	-13738.08	-13774.55	-13868.28
91	-13909.56	-13737.27	-13773.79	-13867.38
92	-15068.76	-14543.52	-14534.28	-14615.17
93	-13974.32	-13801.09	-13844.36	-13930.92
94	-13949.19	-13773.14	-13814.22	-13901.21
95	-13932.00	-13760.12	-13799.45	-13891.28
96	-13909.27	-13736.73	-13773.23	-13866.64
97	-13909.87	-13737.55	-13774.03	-13868.01
98	-14002.83	-13827.89	-13879.58	-13959.12
99	-13938.96	-13760.06	-13798.78	-13889.15
100	-14013.16	-13814.02	-13853.69	-13941.00
All	-1445279.50	-1403987.95	-1403300.05	-1412771.21

- (1a) TI: Thermodynamic integration: log(Prob(D|Model)): Good approximation with many temperatures(1b) BTI: Bezier-approximated Thermodynamic integration: when using few temperatures USE THIS!
- (2) SS: Steppingstone Sampling (Xie et al 2011)
- (3) HS: Harmonic mean approximation: Overestimates the marginal likelihood, poor variance [Scaling factor = 256.452660]

#### Citation suggestions:

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

Palczewski M. and P. Beerli, 2014. Population model comparison using multi-locus datasets. In M.-H. Chen, L. Kuo, and P. O. Lewis, editors, Bayesian Phylogenetics: Methods,

Algorithms, and Applications, pages 187-200. CRC Press, 2014.

Xie W., P. O. Lewis, Y. Fan, L. Kuo, and M.-H. Chen. 2011. Improving marginal likelihood estimation for Bayesian phylogenetic model selection. Systematic Biology, 60(2):150â 160, 2011.

## Acceptance ratios for all parameters and the genealogies

Parameter	Accepted changes	Ratio
$\Theta_1$	232291899/399987119	0.58075
Genealogies	1052990854/1600012881	0.65811

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$ Genealogies	0.07772 0.02814	8641898.67 9498209.02

## Average temperatures during the run

#### Chain Temperatures

- 1 0.00000
- 2 0.00000
- 3 0.00000
- 4 0.00000

Adaptive heating often fails, if the average temperatures are very close together try to rerun using static heating! If you want to compare models using marginal likelihoods then you MUST use static heating

#### Potential Problems

This section reports potential problems with your run, but such reporting is often not very accurate. Whith many parameters in a multilocus analysi s, it is very common that some parameters for some loci will not be very informative, triggering suggestions (for example to increase the prior ran ge) that are not sensible. This suggestion tool will improve with time, therefore do not blindly follow its suggestions. If some parameters are fla

gged, inspect the tables carefully and judge wether an action is required. For example, if you run a Bayesian inference with sequence data, for mac roscopic 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 rou
tes 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