## **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 02:07:37 2017

Program finished at Sun Aug 13 03:48:34 2017 [Runtime:0000:01:40:57]



### **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) 3241463395

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 chains

Recorded steps [a]

Increment (record every x step [b]

Number of concurrent chains (replicates) [c]

1
50000

200

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

bayesallfile\_0.6\_0.5

Print options:

Data file: infile.0.6

Haplotyping is turned on:

Output file: outfile\_0.6\_0.5

Posterior distribution raw histogram file: bayesfile

Print data:

Print genealogies [only some for some data type]:

Raw data from the MCMC run:

## Data summary

Data file: infile.0.6
Datatype: Sequence data
Number of loci: 100

Mutation	model:			
Locus S		Mutationmodel	Mutationmodel parameters	
			·	
1	1	Jukes-Cantor	[Basefreq: =0.25]	
2	1	Jukes-Cantor	[Basefreq: =0.25]	
3	1	Jukes-Cantor	[Basefreq: =0.25]	
4	1	Jukes-Cantor	[Basefreq: =0.25]	
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[Basefreq: =0.25]

Jukes-Cantor

35	1	Jukes-Cantor	[Basefreq: =0.25]
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99	1	Jukes-Cantor	[Basefreq: =0.25]	
100	1	Jukes-Cantor	[Basefreq: =0.25]	
Sites per	locus			
Locus		Sites		
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Locus	Sites
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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	
8	1	1	1.000	1.000	1.000	
9	1	1	1.000	1.000	1.000	
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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	
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33	1	1	1.000	1.000	1.000	
34	1	1	1.000	1.000	1.000	
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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43	1	1	1.000	1.000	1.000	
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45	1	1	1.000	1.000	1.000	
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47	1	1	1.000	1.000	1.000	
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49	1	1	1.000	1.000	1.000	
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52	1	1	1.000	1.000	1.000	
53	1	1	1.000	1.000	1.000	
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56	1	1	1.000	1.000	1.000	
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58	1	1	1.000	1.000	1.000	
59	1	1	1.000	1.000	1.000	
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61	1	1	1.000	1.000	1.000	
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	
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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	
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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	
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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	
94	1	1	1.000	1.000	1.000	
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97	1	1	1.000	1.000	1.000	
98	1	1	1.000	1.000	1.000	
99	1	1	1.000	1.000	1.000	
100	1	1	1.000	1.000	1.000	
Population			1.000	1.000	Locus	Gene copies
1 Romans					1	10
1 Romans	5110111_0				2	10
					3	10
					4	10
					5	10
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	93	10	
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	96	10	
	97	10	
	98	10	
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Total of all populations	1	10	
Total of all populations	2	10	
	3	10	
	4	10	
	5	10	
	6	10	
	7	10	
	8	10	
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95	10
96	10
97	10
98	10
99	10
100	10

# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.02160	0.03700	0.04670	0.04880	0.05093	0.03870	0.05804
2	$\Theta_1$	0.02727	0.04227	0.04770	0.04927	0.05140	0.04243	0.07127
3	$\Theta_1$	0.01653	0.02627	0.03177	0.04420	0.05007	0.03370	0.04607
4	$\Theta_1$	0.02787	0.04193	0.04770	0.04967	0.05147	0.04317	0.07541
5	$\Theta_1$	0.02687	0.04213	0.04763	0.04933	0.05140	0.04250	0.07204
6	$\Theta_1$	0.02000	0.03713	0.04523	0.04840	0.05080	0.03763	0.05607
7	$\Theta_1$	0.02000	0.03693	0.04630	0.04827	0.05067	0.03737	0.05485
8	$\Theta_1$	0.02487	0.03973	0.04763	0.04933	0.05120	0.04117	0.06621
9	$\Theta_1$	0.01587	0.02813	0.03397	0.04127	0.05027	0.03370	0.04608
10	$\Theta_1$	0.02433	0.03960	0.04757	0.04920	0.05120	0.04083	0.06604
11	$\Theta_1$	0.02407	0.03927	0.04757	0.04927	0.05120	0.04070	0.06537
12	$\Theta_1$	0.02667	0.04087	0.04763	0.04933	0.05133	0.04223	0.07186
13	$\Theta_1$	0.02333	0.03947	0.04757	0.04907	0.05113	0.04003	0.06211
14	$\Theta_1$	0.02113	0.03640	0.04750	0.04860	0.05080	0.03810	0.05585
15	$\Theta_1$	0.02107	0.03820	0.04757	0.04867	0.05093	0.03843	0.05880
16	$\Theta_1$	0.02113	0.03673	0.04750	0.04873	0.05087	0.03830	0.05716
17	$\Theta_1$	0.02473	0.04093	0.04763	0.04920	0.05127	0.04117	0.06761
18	$\Theta_1$	0.02153	0.03767	0.04757	0.04867	0.05093	0.03857	0.05767
								-

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 02:07:37]

19	$\Theta_1$	0.02127	0.03727	0.04417	0.04847	0.05080	0.03817	0.05651
20	$\Theta_1$	0.02060	0.03720	0.04637	0.04853	0.05080	0.03810	0.05625
21	$\Theta_1$	0.02433	0.04080	0.04770	0.04927	0.05127	0.04117	0.07038
22	$\Theta_1$	0.02960	0.04267	0.04770	0.04953	0.05153	0.04397	0.07805
23	$\Theta_1$	0.02187	0.03800	0.04750	0.04893	0.05107	0.03930	0.06105
24	$\Theta_1$	0.02133	0.03313	0.04643	0.04927	0.05093	0.03863	0.05722
25	$\Theta_1$	0.02480	0.04080	0.04757	0.04907	0.05120	0.04110	0.06545
26	$\Theta_1$	0.02853	0.04293	0.04770	0.04940	0.05140	0.04310	0.07471
27	$\Theta_1$	0.01713	0.03207	0.03723	0.04640	0.05047	0.03530	0.04968
28	$\Theta_1$	0.02387	0.04027	0.04757	0.04900	0.05113	0.04043	0.06449
29	$\Theta_1$	0.02233	0.03873	0.04750	0.04867	0.05087	0.03910	0.05875
30	$\Theta_1$	0.02393	0.03893	0.04757	0.04913	0.05113	0.04043	0.06379
31	$\Theta_1$	0.02400	0.03953	0.04763	0.04933	0.05127	0.04090	0.06640
32	$\Theta_1$	0.01907	0.03553	0.04263	0.04787	0.05060	0.03650	0.05270
33	$\Theta_1$	0.02453	0.03987	0.04763	0.04933	0.05133	0.04123	0.06716
34	$\Theta_1$	0.01933	0.03660	0.04070	0.04773	0.05067	0.03683	0.05283
35	$\Theta_1$	0.02313	0.03967	0.04750	0.04880	0.05107	0.03983	0.06151
36	$\Theta_1$	0.02253	0.03807	0.04717	0.04887	0.05093	0.03950	0.05972
37	$\Theta_1$	0.01633	0.02553	0.03430	0.04533	0.05000	0.03370	0.04609
38	$\Theta_1$	0.02660	0.04213	0.04763	0.04940	0.05133	0.04230	0.07019
39	$\Theta_1$	0.01633	0.02693	0.03537	0.04293	0.05007	0.03370	0.04607
40	$\Theta_1$	0.01780	0.03153	0.04130	0.04567	0.05033	0.03523	0.04938
41	$\Theta_1$	0.02180	0.03733	0.04750	0.04887	0.05100	0.03897	0.05851

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.02313	0.03933	0.04763	0.04933	0.05127	0.04077	0.06817
43	$\Theta_1$	0.02593	0.04173	0.04757	0.04933	0.05127	0.04210	0.07219
44	$\Theta_1$	0.02000	0.03613	0.04750	0.04833	0.05080	0.03743	0.05431
45	$\Theta_1$	0.02213	0.02600	0.04657	0.05040	0.05087	0.03897	0.05857
46	$\Theta_1$	0.01913	0.03460	0.03883	0.04813	0.05060	0.03663	0.05310
47	$\Theta_1$	0.01780	0.03147	0.03750	0.04620	0.05040	0.03523	0.04962
48	$\Theta_1$	0.02580	0.04020	0.04763	0.04927	0.05127	0.04170	0.06859
49	$\Theta_1$	0.01800	0.03240	0.03830	0.04493	0.05033	0.03543	0.04998
50	$\Theta_1$	0.02407	0.03960	0.04757	0.04900	0.05120	0.04050	0.06324
51	$\Theta_1$	0.01960	0.03580	0.04137	0.04727	0.05073	0.03697	0.05301
52	$\Theta_1$	0.02213	0.03907	0.04750	0.04873	0.05100	0.03923	0.06055
53	$\Theta_1$	0.02707	0.04140	0.04770	0.04960	0.05147	0.04263	0.07178
54	$\Theta_1$	0.02627	0.04147	0.04763	0.04960	0.05147	0.04270	0.07386
55	$\Theta_1$	0.01987	0.03713	0.04717	0.04840	0.05080	0.03757	0.05599
56	$\Theta_1$	0.02020	0.02833	0.04197	0.04953	0.05073	0.03757	0.05444
57	$\Theta_1$	0.02827	0.04207	0.04770	0.04953	0.05147	0.04337	0.07810
58	$\Theta_1$	0.02320	0.03900	0.04750	0.04900	0.05107	0.03997	0.06230
59	$\Theta_1$	0.02220	0.03747	0.04750	0.04887	0.05093	0.03903	0.05915
60	$\Theta_1$	0.01727	0.03187	0.03730	0.04487	0.05040	0.03523	0.04921
61	$\Theta_1$	0.01773	0.03107	0.03990	0.04540	0.05040	0.03523	0.04952

62	$\Theta_1$	0.02453	0.04060	0.04757	0.04907	0.05120	0.04077	0.06429
63	$\Theta_1$	0.02133	0.03827	0.04750	0.04860	0.05087	0.03843	0.05802
64	$\Theta_1$	0.01973	0.03707	0.04630	0.04807	0.05067	0.03723	0.05374
65	$\Theta_1$	0.02320	0.03880	0.04757	0.04913	0.05120	0.04030	0.06506
66	$\Theta_1$	0.02447	0.04000	0.04763	0.04933	0.05133	0.04143	0.06993
67	$\Theta_1$	0.02900	0.04347	0.04777	0.04933	0.05140	0.04363	0.07715
68	$\Theta_1$	0.02727	0.04133	0.04763	0.04953	0.05140	0.04263	0.07105
69	$\Theta_1$	0.01840	0.03267	0.04230	0.04633	0.05047	0.03583	0.05100
70	$\Theta_1$	0.01833	0.03667	0.03990	0.04800	0.05100	0.03703	0.05298
71	$\Theta_1$	0.02100	0.03787	0.04237	0.04833	0.05080	0.03803	0.05571
72	$\Theta_1$	0.02307	0.03913	0.04750	0.04920	0.05120	0.04057	0.06512
73	$\Theta_1$	0.02553	0.04140	0.04763	0.04907	0.05127	0.04163	0.06778
74	$\Theta_1$	0.02753	0.04167	0.04770	0.04953	0.05140	0.04290	0.07333
75	$\Theta_1$	0.02653	0.04200	0.04763	0.04927	0.05133	0.04217	0.07203
76	$\Theta_1$	0.02880	0.04213	0.04783	0.04967	0.05147	0.04337	0.07564
77	$\Theta_1$	0.02033	0.03740	0.04603	0.04813	0.05073	0.03763	0.05484
78	$\Theta_1$	0.01787	0.03067	0.03717	0.04447	0.05033	0.03517	0.04933
79	$\Theta_1$	0.02160	0.03740	0.04750	0.04893	0.05107	0.03903	0.05929
80	$\Theta_1$	0.02300	0.03953	0.04757	0.04913	0.05113	0.04017	0.06343
81	$\Theta_1$	0.01967	0.03713	0.04350	0.04827	0.05080	0.03743	0.05495
82	$\Theta_1$	0.02527	0.04000	0.04763	0.04940	0.05127	0.04137	0.06849
83	$\Theta_1$	0.01867	0.03140	0.03670	0.04807	0.05053	0.03597	0.05064
84	$\Theta_1$	0.02547	0.04120	0.04763	0.04927	0.05127	0.04157	0.06806

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.02140	0.03693	0.04750	0.04873	0.05087	0.03843	0.05798
86	$\Theta_1$	0.02660	0.04087	0.04763	0.04947	0.05133	0.04223	0.07121
87	$\Theta_1$	0.01880	0.03387	0.04170	0.04753	0.05047	0.03623	0.05118
88	$\Theta_1$	0.02287	0.03947	0.04750	0.04880	0.05107	0.03970	0.06074
89	$\Theta_1$	0.02260	0.03840	0.04757	0.04893	0.05100	0.03950	0.06081
90	$\Theta_1$	0.02353	0.03887	0.04757	0.04920	0.05120	0.04037	0.06416
91	$\Theta_1$	0.02220	0.03820	0.04757	0.04887	0.05100	0.03923	0.05951
92	$\Theta_1$	0.02473	0.04100	0.04770	0.04920	0.05127	0.04123	0.06928
93	$\Theta_1$	0.02513	0.04107	0.04763	0.04927	0.05127	0.04143	0.06797
94	$\Theta_1$	0.02327	0.03867	0.04737	0.04907	0.05107	0.04017	0.06327
95	$\Theta_1$	0.02380	0.03993	0.04750	0.04887	0.05100	0.04017	0.06191
96	$\Theta_1$	0.03000	0.04253	0.04777	0.04960	0.05147	0.04383	0.07806
97	$\Theta_1$	0.02553	0.04053	0.04763	0.04933	0.05127	0.04170	0.06841
98	$\Theta_1$	0.02620	0.04040	0.04763	0.04933	0.05127	0.04177	0.06744
99	$\Theta_1$	0.01967	0.03673	0.04430	0.04813	0.05067	0.03723	0.05338
100	$\Theta_1$	0.02567	0.04180	0.04763	0.04927	0.05140	0.04197	0.07157
All	$\Theta_1$	0.04680	0.04833	0.04937	0.05033	0.05187	0.04943	0.05839

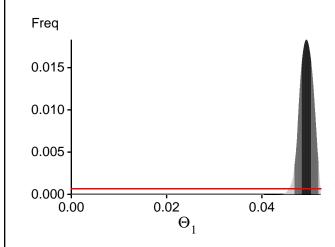
Citation suggestions:

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

ocus.	TI(1a)	BTI(1b)	SS(2)	HS(3)
1	-13930.60	-13777.32	-13825.31	-13911.50
2	-14762.76	-14406.87	-14436.74	-14510.48
3	-13873.89	-13729.86	-13773.76	-13866.34
4	-14282.46	-14094.61	-14153.73	-14232.04
5	-14450.97	-14186.99	-14231.80	-14305.64
6	-13908.98	-13764.55	-13815.16	-13901.57
7	-13912.25	-13763.95	-13812.47	-13900.18
8	-14213.58	-13970.35	-14011.68	-14091.11
9	-13869.15	-13725.12	-13768.72	-13861.70
10	-14569.31	-14240.35	-14268.09	-14348.11
11	-13986.97	-13823.52	-13877.82	-13957.05
12	-14453.34	-14172.70	-14213.00	-14289.42
13	-13959.32	-13806.99	-13858.12	-13942.44
14	-13956.04	-13798.73	-13847.79	-13933.17
15	-13924.46	-13778.33	-13829.81	-13915.39
16	-13932.67	-13777.81	-13827.57	-13912.83
17	-16342.14	-15760.93	-15759.88	-15837.35
18	-14250.58	-14014.56	-14052.05	-14136.68
19	-13948.89	-13784.41	-13831.71	-13917.72
20	-13916.89	-13768.35	-13816.19	-13902.92
21	-14666.68	-14449.13	-14485.03	-14572.90
22	-17435.37	-15734.01	-15523.03	-15592.61
23	-13939.03	-13790.82	-13842.16	-13926.20
24	-14009.34	-13829.93	-13875.74	-13960.72
25	-14042.10	-13863.64	-13914.52	-13995.34
26	-15399.38	-14783.87	-14771.48	-14841.85
27	-13887.80	-13742.17	-13786.45	-13877.78
28	-17450.96	-15688.78	-15456.14	-15536.12
29	-13991.39	-13816.32	-13863.65	-13947.55

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 02:07:37]

30	-13978.79	-13823.93	-13876.81	-13958.70
31	-13993.88	-13836.16	-13890.11	-13976.86
32	-13898.55	-13754.11	-13802.01	-13890.96
33	-14010.35	-13858.37	-13915.30	-13993.91
34	-13934.34	-13786.70	-13834.43	-13922.95
35	-14268.32	-14082.55	-14133.81	-14215.50
36	-14178.98	-13942.65	-13980.59	-14064.63
37	-13873.12	-13728.80	-13772.69	-13865.40
38	-14273.97	-14033.71	-14078.94	-14154.71
39	-13873.53	-13729.44	-13773.36	-13865.91
40	-13886.02	-13741.97	-13788.23	-13878.88
41	-13968.58	-13800.25	-13848.08	-13932.58
42	-27353.12	-22928.60	-22263.04	-22355.75
43	-14360.24	-14150.34	-14200.37	-14278.22
44	-13924.47	-13773.90	-13822.24	-13909.64
45	-13985.93	-13813.38	-13862.26	-13946.02
46	-13903.94	-13755.68	-13801.84	-13892.31
47	-13887.42	-13742.95	-13788.87	-13879.43
48	-14103.09	-13912.36	-13963.97	-14043.12
49	-13889.03	-13741.82	-13785.80	-13879.13
50	-14035.88	-13858.47	-13908.44	-13990.80
51	-13939.21	-13791.02	-13839.91	-13927.09
52	-13937.94	-13791.36	-13842.31	-13926.60
53	-14182.90	-13973.00	-14024.88	-14099.61
54	-14758.89	-14485.91	-14531.22	-14606.07
55	-13910.76	-13766.26	-13816.21	-13903.44
56	-14016.03	-13830.90	-13874.20	-13960.31
57	-16699.29	-16279.51	-16286.65	-16369.30
58	-13995.78	-13824.09	-13874.01	-13956.41
59	-13956.36	-13797.75	-13847.11	-13931.87
60	-13886.31	-13742.05	-13784.62	-13878.65
61	-13890.62	-13742.75	-13786.69	-13880.92
62	-13993.68	-13834.04	-13887.43	-13968.83
63	-13936.70	-13786.23	-13836.85	-13921.87
64	-13919.59	-13764.58	-13811.82	-13900.25
65	-28324.37	-20976.37	-19732.90	-19810.18
66	-14126.23	-13954.07	-14005.12	-14086.20
67	-15120.03	-14585.52	-14586.29	-14655.54
68	-14209.44	-13970.17	-14016.42	-14090.57
69	-13894.92	-13750.26	-13795.38	-13887.25
70	-13970.50	-13800.34	-13845.06	-13932.72
71	-13966.14	-13800.41	-13848.43	-13933.61
72	-15009.50	-14559.32	-14567.46	-14646.96
73	-14412.34	-14075.66	-14100.37	-14178.37
74	-14112.56	-13929.95	-13986.46	-14060.35

75	-14090.91	-13923.38	-13977.32	-14056.77
76	-14321.40	-14110.30	-14164.35	-14237.32
77	-13942.87	-13784.25	-13832.17	-13918.41
78	-13885.84	-13737.96	-13781.98	-13876.03
79	-13941.65	-13788.22	-13839.48	-13924.06
80	-13970.92	-13811.20	-13863.67	-13945.30
81	-13914.15	-13767.90	-13815.55	-13904.04
82	-20273.60	-18606.35	-18430.67	-18506.29
83	-13904.96	-13753.90	-13799.16	-13889.29
84	-14049.08	-13892.18	-13949.39	-14027.63
85	-13939.60	-13788.91	-13839.52	-13924.80
86	-14080.55	-13909.07	-13965.11	-14041.63
87	-13906.39	-13761.22	-13807.13	-13897.56
88	-13992.95	-13820.63	-13869.59	-13954.24
89	-13968.91	-13811.90	-13863.43	-13946.12
90	-13979.72	-13829.90	-13882.62	-13965.65
91	-13960.01	-13797.04	-13847.82	-13930.14
92	-14976.58	-14655.23	-14690.14	-14768.05
93	-13998.87	-13845.18	-13900.83	-13978.58
94	-13990.23	-13823.19	-13874.01	-13956.73
95	-14130.17	-13914.03	-13955.89	-14038.11
96	-18228.94	-16305.67	-16061.44	-16129.22
97	-15531.86	-14706.17	-14645.08	-14723.12
98	-14356.63	-14047.29	-14078.14	-14155.07
99	-13915.66	-13764.55	-13812.41	-13899.76
100	-16374.89	-15496.91	-15436.03	-15515.19
All	-1462616.54	-1423875.36	-1425101.76	-1433412.42

- (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 = 29.746952]

#### 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
$\Theta_1$ Genealogies	384505380/399995576 516782348/1600004424	0.96127 0.32299

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.67905	1915043.34
Genealogies	0.09085	8572477.33

## 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