## **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 11:14:38 2017

Program finished at Sun Aug 13 12:35:19 2017 [Runtime:0000:01:20:41]



## **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) 949156198

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.0.8

Haplotyping is turned on:

Output file: outfile\_0.8\_0.5

Posterior distribution raw histogram file: bayesfile

Raw data from the MCMC run: bayesallfile\_0.8\_0.5
Print data: No

Print genealogies [only some for some data type]:

## Data summary

Data file: infile.0.8
Datatype: Sequence data
Number of loci: 100

T TOTTIBET T	01 1001.			100
Mutation	model:			
Locus Su	ublocus	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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32	1	Jukes-Cantor	[Basefreq: =0.25]	
33	1	Jukes-Cantor	[Basefreq: =0.25]	

[Basefreq: =0.25]

Jukes-Cantor

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35	1	Jukes-Cantor	[Pagefreg: -0.25]
36	1 1	Jukes-Cantor	[Basefreq: =0.25] [Basefreq: =0.25]
37	1	Jukes-Cantor	[Basefreq: =0.25]
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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	
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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	
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	
38	1	1	1.000	1.000	1.000	
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41	1	1	1.000	1.000	1.000	
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	

52	1	1	1.000	1.000	1.000	
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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	
60	1	1	1.000	1.000	1.000	
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	
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	
94	1	1	1.000	1.000	1.000	
95	1	1	1.000	1.000	1.000	
96	1	1	1.000	1.000	1.000	

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	
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Total of all populations	1	10	
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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	$\Theta_1$	0.03180	0.04447	0.04790	0.04953	0.05153	0.04497	0.08258
2	$\Theta_1$	0.02773	0.04160	0.04770	0.04953	0.05147	0.04290	0.07348
3	$\Theta_1$	0.03147	0.04427	0.04777	0.04940	0.05147	0.04443	0.08039
4	$\Theta_1$	0.02773	0.04260	0.04777	0.04940	0.05140	0.04277	0.07409
5	$\Theta_1$	0.03107	0.04360	0.04790	0.04973	0.05160	0.04477	0.08143
6	$\Theta_1$	0.03160	0.04327	0.04777	0.04973	0.05153	0.04450	0.08033
7	$\Theta_1$	0.02993	0.04380	0.04777	0.04940	0.05153	0.04397	0.07702
8	$\Theta_1$	0.02947	0.04267	0.04770	0.04967	0.05147	0.04390	0.07700
9	$\Theta_1$	0.02807	0.04280	0.04763	0.04927	0.05140	0.04297	0.07454
10	$\Theta_1$	0.02787	0.04267	0.04763	0.04933	0.05140	0.04283	0.07313
11	$\Theta_1$	0.03033	0.04340	0.04790	0.04987	0.05160	0.04457	0.08259
12	$\Theta_1$	0.03073	0.04093	0.04763	0.05000	0.05153	0.04423	0.07899
13	$\Theta_1$	0.02900	0.04280	0.04770	0.04967	0.05153	0.04397	0.07927
14	$\Theta_1$	0.03207	0.04353	0.04783	0.04960	0.05160	0.04477	0.08123
15	$\Theta_1$	0.02987	0.04287	0.04777	0.04967	0.05153	0.04410	0.07968
16	$\Theta_1$	0.03180	0.04353	0.04783	0.04973	0.05153	0.04477	0.08134
17	$\Theta_1$	0.02880	0.04253	0.04777	0.04967	0.05147	0.04370	0.07718
18	$\Theta_1$	0.03120	0.04367	0.04790	0.04987	0.05160	0.04483	0.08040

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 11:14:38]

19	$\Theta_1$	0.03340	0.04413	0.04790	0.05000	0.05153	0.04523	0.08448
20	$\Theta_1$	0.03193	0.04420	0.04777	0.04953	0.05153	0.04483	0.08187
21	$\Theta_1$	0.02993	0.04273	0.04777	0.04960	0.05147	0.04403	0.07829
22	$\Theta_1$	0.02667	0.04100	0.04763	0.04940	0.05133	0.04243	0.07086
23	$\Theta_1$	0.03287	0.04387	0.04783	0.04973	0.05153	0.04503	0.08227
24	$\Theta_1$	0.03093	0.04333	0.04777	0.04967	0.05160	0.04457	0.08077
25	$\Theta_1$	0.03227	0.04433	0.04790	0.04987	0.05167	0.04543	0.08395
26	$\Theta_1$	0.03187	0.04367	0.04797	0.04980	0.05167	0.04483	0.08273
27	$\Theta_1$	0.03113	0.04327	0.04783	0.04980	0.05153	0.04443	0.08061
28	$\Theta_1$	0.02867	0.04327	0.04770	0.04947	0.05147	0.04343	0.07676
29	$\Theta_1$	0.03160	0.04387	0.04790	0.04980	0.05160	0.04503	0.08377
30	$\Theta_1$	0.02753	0.04253	0.04770	0.04933	0.05140	0.04277	0.07273
31	$\Theta_1$	0.03227	0.04380	0.04783	0.04987	0.05153	0.04490	0.08180
32	$\Theta_1$	0.02953	0.04253	0.04770	0.04967	0.05147	0.04377	0.07699
33	$\Theta_1$	0.03393	0.04467	0.04790	0.04980	0.05167	0.04583	0.08554
34	$\Theta_1$	0.03080	0.04313	0.04777	0.04967	0.05160	0.04437	0.08207
35	$\Theta_1$	0.02687	0.04207	0.04770	0.04940	0.05133	0.04237	0.07270
36	$\Theta_1$	0.02813	0.04187	0.04770	0.04960	0.05147	0.04310	0.07437
37	$\Theta_1$	0.03080	0.04427	0.04783	0.04960	0.05167	0.04443	0.08142
38	$\Theta_1$	0.02767	0.04260	0.04763	0.04927	0.05133	0.04277	0.07095
39	$\Theta_1$	0.02960	0.04313	0.04777	0.04967	0.05153	0.04383	0.07994
40	$\Theta_1$	0.02733	0.04140	0.04770	0.04953	0.05153	0.04277	0.07309
41	$\Theta_1$	0.02767	0.04180	0.04763	0.04953	0.05147	0.04310	0.07516

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03113	0.04367	0.04777	0.04973	0.05160	0.04483	0.08286
43	$\Theta_1$	0.03260	0.04400	0.04783	0.04967	0.05160	0.04517	0.08329
44	$\Theta_1$	0.02973	0.04360	0.04770	0.04933	0.05147	0.04377	0.07832
45	$\Theta_1$	0.03127	0.04313	0.04783	0.04967	0.05160	0.04437	0.08002
46	$\Theta_1$	0.03160	0.04333	0.04797	0.04980	0.05153	0.04450	0.08242
47	$\Theta_1$	0.03100	0.04347	0.04770	0.04960	0.05160	0.04470	0.08021
48	$\Theta_1$	0.02633	0.04100	0.04763	0.04953	0.05140	0.04230	0.07331
49	$\Theta_1$	0.03113	0.04333	0.04783	0.04980	0.05153	0.04450	0.08164
50	$\Theta_1$	0.02820	0.04193	0.04777	0.04960	0.05147	0.04317	0.07372
51	$\Theta_1$	0.03413	0.04427	0.04783	0.04960	0.05153	0.04557	0.08664
52	$\Theta_1$	0.03273	0.04513	0.04783	0.04953	0.05173	0.04550	0.08550
53	$\Theta_1$	0.03107	0.04347	0.04783	0.04973	0.05153	0.04470	0.08084
54	$\Theta_1$	0.03100	0.04333	0.04777	0.04967	0.05153	0.04450	0.08043
55	$\Theta_1$	0.02987	0.04260	0.04777	0.04967	0.05160	0.04383	0.07977
56	$\Theta_1$	0.02800	0.03867	0.04770	0.05013	0.05140	0.04310	0.07328
57	$\Theta_1$	0.02927	0.04227	0.04770	0.04960	0.05153	0.04350	0.07806
58	$\Theta_1$	0.03000	0.04300	0.04783	0.04980	0.05167	0.04417	0.08033
59	$\Theta_1$	0.02040	0.03653	0.04670	0.04833	0.05073	0.03763	0.05456
60	$\Theta_1$	0.02967	0.04280	0.04777	0.04973	0.05160	0.04403	0.07847
61	$\Theta_1$	0.03427	0.04427	0.04797	0.04980	0.05160	0.04550	0.08501

62	$\Theta_1$	0.03187	0.04113	0.04777	0.05013	0.05153	0.04463	0.08184
63	$\Theta_1$	0.03060	0.04320	0.04790	0.04980	0.05160	0.04443	0.08037
64	$\Theta_1$	0.02453	0.03980	0.04757	0.04927	0.05127	0.04123	0.06566
65	$\Theta_1$	0.02980	0.04300	0.04777	0.04967	0.05167	0.04423	0.07968
66	$\Theta_1$	0.02747	0.04180	0.04763	0.04953	0.05147	0.04310	0.07404
67	$\Theta_1$	0.03093	0.04347	0.04790	0.04980	0.05160	0.04463	0.08270
68	$\Theta_1$	0.03093	0.04360	0.04777	0.04973	0.05160	0.04483	0.08046
69	$\Theta_1$	0.02967	0.04287	0.04777	0.04960	0.05153	0.04410	0.07937
70	$\Theta_1$	0.02720	0.04233	0.04770	0.04933	0.05133	0.04250	0.07183
71	$\Theta_1$	0.03293	0.04427	0.04790	0.04980	0.05173	0.04543	0.08315
72	$\Theta_1$	0.02813	0.04293	0.04777	0.04940	0.05140	0.04310	0.07246
73	$\Theta_1$	0.02947	0.04247	0.04777	0.04967	0.05140	0.04370	0.07910
74	$\Theta_1$	0.03260	0.04400	0.04797	0.04993	0.05167	0.04510	0.08350
75	$\Theta_1$	0.03260	0.04380	0.04790	0.04973	0.05160	0.04503	0.08242
76	$\Theta_1$	0.02700	0.04120	0.04763	0.04947	0.05140	0.04250	0.07192
77	$\Theta_1$	0.03380	0.04447	0.04777	0.04967	0.05160	0.04563	0.08479
78	$\Theta_1$	0.03167	0.04353	0.04790	0.04987	0.05160	0.04470	0.08128
79	$\Theta_1$	0.02853	0.04180	0.04770	0.04953	0.05147	0.04310	0.07321
80	$\Theta_1$	0.03073	0.04333	0.04783	0.04973	0.05167	0.04457	0.08038
81	$\Theta_1$	0.02900	0.04227	0.04777	0.04960	0.05140	0.04350	0.07762
82	$\Theta_1$	0.02953	0.04387	0.04783	0.04953	0.05147	0.04403	0.07853
83	$\Theta_1$	0.03060	0.04347	0.04790	0.04993	0.05167	0.04457	0.08241
84	$\Theta_1$	0.03140	0.04327	0.04783	0.04967	0.05167	0.04457	0.08267

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.03073	0.04307	0.04777	0.04967	0.05153	0.04430	0.07854
86	$\Theta_1$	0.02973	0.04293	0.04777	0.04967	0.05153	0.04417	0.07906
87	$\Theta_1$	0.03233	0.04400	0.04777	0.04973	0.05153	0.04517	0.08302
88	$\Theta_1$	0.02933	0.04260	0.04777	0.04967	0.05153	0.04390	0.07805
89	$\Theta_1$	0.03087	0.04333	0.04777	0.04973	0.05160	0.04457	0.08033
90	$\Theta_1$	0.02840	0.04220	0.04777	0.04967	0.05153	0.04343	0.07733
91	$\Theta_1$	0.03267	0.04420	0.04790	0.04987	0.05160	0.04537	0.08329
92	$\Theta_1$	0.03173	0.04380	0.04770	0.04973	0.05167	0.04497	0.08250
93	$\Theta_1$	0.02933	0.04267	0.04770	0.04960	0.05153	0.04397	0.07869
94	$\Theta_1$	0.02767	0.04260	0.04770	0.04927	0.05140	0.04277	0.07366
95	$\Theta_1$	0.02633	0.04187	0.04763	0.04933	0.05140	0.04243	0.07197
96	$\Theta_1$	0.02873	0.04227	0.04770	0.04967	0.05147	0.04350	0.07623
97	$\Theta_1$	0.03073	0.04327	0.04770	0.04960	0.05153	0.04430	0.07969
98	$\Theta_1$	0.03067	0.04313	0.04777	0.04973	0.05153	0.04437	0.08130
99	$\Theta_1$	0.03047	0.04300	0.04783	0.04973	0.05153	0.04423	0.07876
100	$\Theta_1$	0.02967	0.04287	0.04783	0.04973	0.05153	0.04410	0.07942
All	$\Theta_1$	0.00687	0.00880	0.01030	0.01147	0.01393	0.01037	0.09931

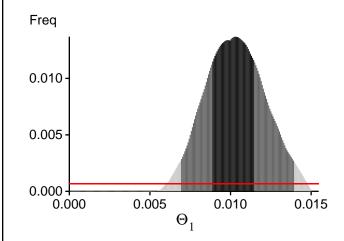
#### Citation suggestions:

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	-15223.87	-14894.64	-14944.44	-15008.26
2	-14180.03	-13973.56	-14024.95	-14099.65
3	-14550.20	-14236.92	-14275.95	-14344.71
4	-15538.24	-15113.36	-15139.80	-15212.47
5	-14663.43	-14356.43	-14405.04	-14470.50
6	-15350.66	-14807.15	-14814.00	-14878.82
7	-14597.33	-14293.66	-14335.78	-14406.45
8	-14277.11	-14091.77	-14151.09	-14222.80
9	-14127.35	-13944.59	-14002.83	-14075.42
10	-14379.39	-14105.04	-14146.61	-14220.67
11	-16346.58	-15700.90	-15698.18	-15758.50
12	-14801.93	-14532.63	-14584.96	-14654.24
13	-14719.28	-14424.98	-14470.88	-14539.63
14	-14607.03	-14368.84	-14428.13	-14493.35
15	-15201.41	-14802.54	-14837.23	-14901.86
16	-16300.00	-15289.18	-15209.01	-15275.61
17	-14945.48	-14681.80	-14729.80	-14802.48
18	-14370.06	-14146.14	-14204.92	-14270.88
19	-16282.92	-15469.21	-15429.66	-15493.08
20	-14669.82	-14360.75	-14408.65	-14471.77
21	-16275.95	-15495.38	-15464.84	-15529.98
22	-14078.60	-13903.42	-13960.26	-14035.67
23	-14987.77	-14524.04	-14544.87	-14606.59
24	-14495.52	-14280.92	-14338.94	-14405.55
25	-15347.68	-14748.55	-14741.87	-14807.70
26	-15443.63	-14916.92	-14929.31	-14992.20
27	-15432.33	-14771.78	-14750.99	-14818.44
28	-14547.88	-14253.04	-14297.82	-14366.24
29	-36244.49	-24422.57	-22359.54	-22443.03

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 11:14:38]

30	-15609.29	-14760.67	-14699.08	-14773.11
31	-14600.47	-14306.58	-14357.39	-14420.21
32	-14477.37	-14185.72	-14223.82	-14296.26
33	-19390.77	-18116.64	-18028.14	-18084.24
34	-20960.30	-18817.11	-18568.39	-18627.89
35	-14060.12	-13900.57	-13958.57	-14034.06
36	-14433.05	-14138.88	-14177.52	-14251.22
37	-16784.89	-16018.16	-15997.50	-16059.74
38	-14264.38	-14042.62	-14092.47	-14167.80
39	-15167.75	-14789.28	-14827.24	-14891.30
40	-14100.88	-13924.54	-13981.25	-14055.12
41	-15171.42	-14724.80	-14741.13	-14814.33
42	-15035.42	-14644.78	-14678.96	-14744.65
43	-15229.99	-14791.74	-14816.01	-14879.96
44	-14312.64	-14105.78	-14165.94	-14233.70
45	-15209.76	-14746.00	-14768.91	-14832.34
46	-15898.76	-15063.10	-15016.20	-15080.02
47	-15008.70	-14478.34	-14480.83	-14548.48
48	-14110.02	-13953.20	-14009.80	-14086.12
49	-14396.45	-14189.53	-14253.16	-14315.93
50	-14576.69	-14339.39	-14391.54	-14464.63
51	-49967.05	-32076.62	-28950.88	-29009.02
52	-22869.38	-19599.25	-19150.99	-19201.51
53	-14625.89	-14282.40	-14320.01	-14385.99
54	-14339.59	-14133.59	-14194.81	-14260.86
55	-27628.76	-23838.65	-23329.20	-23392.24
56	-15373.40	-14637.25	-14596.90	-14670.57
57	-14510.59	-14282.95	-14332.49	-14404.46
58	-15150.37	-14747.57	-14777.10	-14844.51
59	-13935.10	-13775.45	-13822.47	-13908.52
60	-14300.40	-14098.99	-14155.28	-14225.15
61	-15756.66	-14965.10	-14925.46	-14990.05
62	-14623.79	-14324.17	-14369.83	-14433.79
63	-14614.12	-14318.04	-14362.24	-14430.05
64	-14073.60	-13897.74	-13951.18	-14029.94
65	-14369.57	-14122.63	-14176.00	-14242.45
66	-14116.47	-13927.97	-13983.11	-14059.01
67	-14696.14	-14376.53	-14423.98	-14486.73
68	-14822.31	-14450.27	-14481.92	-14549.32
69	-14683.25	-14365.58	-14405.83	-14476.27
70	-14098.07	-13923.31	-13980.81	-14055.45
71	-15789.45	-15063.89	-15042.64	-15104.07
72	-14335.20	-14056.60	-14096.20	-14172.83
73	-14488.97	-14237.58	-14289.03	-14359.06
74	-15071.31	-14613.06	-14638.04	-14698.47

75	-14653.17	-14367.84	-14419.22	-14482.89
76	-14058.49	-13899.52	-13957.04	-14032.92
77	-15053.98	-14729.07	-14782.52	-14839.61
78	-15632.45	-14854.80	-14816.36	-14881.24
79	-14306.26	-14040.98	-14084.55	-14158.48
80	-14532.64	-14233.96	-14280.32	-14346.64
81	-14182.80	-13996.61	-14055.90	-14127.43
82	-14302.30	-14091.09	-14146.70	-14216.86
83	-17104.37	-15962.63	-15870.95	-15932.36
84	-15073.98	-14770.72	-14823.25	-14884.75
85	-14499.78	-14192.08	-14235.40	-14301.76
86	-14750.00	-14332.06	-14353.36	-14422.34
87	-15640.26	-15192.03	-15223.20	-15286.10
88	-14351.94	-14163.42	-14224.72	-14295.03
89	-14537.21	-14269.09	-14319.49	-14390.75
90	-14187.01	-14000.51	-14061.33	-14130.19
91	-15281.68	-14742.14	-14753.85	-14814.11
92	-14945.06	-14580.64	-14617.62	-14681.06
93	-26847.77	-22850.12	-22256.67	-22356.36
94	-14596.28	-14227.35	-14247.82	-14324.66
95	-14801.86	-14444.87	-14475.56	-14549.43
96	-14584.47	-14237.31	-14269.61	-14340.84
97	-14790.87	-14454.83	-14494.78	-14562.63
98	-14647.04	-14336.07	-14380.62	-14446.18
99	-14489.49	-14196.27	-14237.31	-14309.27
100	-19803.67	-17442.00	-17130.64	-17195.94
All	-1588579.71	-1507571.63	-1503039.69	-1509868.15

- (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 = 99.672319]

#### 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	382939117/399972646 178345043/1600027354	0.95741 0.11146

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$ Genealogies	0.56577 0.25980	2787204.41 6041672.17

## Average temperatures during the run

# Chain Temperatures 1 0.00000

2 0.00000

3 0.000004 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

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