## **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 40 compute nodes are available.

Program started at Mon Jul 24 14:46:07 2017

Program finished at Mon Jul 24 18:45:17 2017 [Runtime:0000:03:59:10]



### **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) 1902430322

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\_0.8\_0.6

Print options:

Data file: infile.0.8

Haplotyping is turned on:

Output file: outfile\_0.8\_0.6

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.0.8
Datatype: Sequence data
Number of loci: 100

Mutationmodel	•
Mutationinous	

Mutatio	nmodel:			
Locus	Sublocus	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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100	1	Jukes-Cantor	[Basefreq: =0.25]	
Sites per	locus			
Locus		Sites		

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	
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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	
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31	1	1	1.000	1.000	1.000	
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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	
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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	
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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	
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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	
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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	1.000	1.000	Locus	Gene copies
1 Romans					1	10
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Total of all populations	1	10	
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	99	10
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# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.02560	0.03987	0.04757	0.04920	0.05120	0.04130	0.06476
2	$\Theta_1$	0.01607	0.02807	0.03543	0.04353	0.05007	0.03377	0.04734
3	$\Theta_1$	0.02213	0.03900	0.04757	0.04893	0.05100	0.03943	0.06280
4	$\Theta_1$	0.02247	0.03853	0.04750	0.04887	0.05100	0.03943	0.05978
5	$\Theta_1$	0.01700	0.02727	0.03617	0.04360	0.05020	0.03423	0.04523
6	$\Theta_1$	0.02160	0.03747	0.04597	0.04860	0.05087	0.03863	0.05739
7	$\Theta_1$	0.01627	0.02787	0.03343	0.04447	0.05020	0.03397	0.04766
8	$\Theta_1$	0.02313	0.03947	0.04710	0.04880	0.05107	0.03977	0.05972
9	$\Theta_1$	0.02167	0.03767	0.04570	0.04833	0.05087	0.03857	0.05543
10	$\Theta_1$	0.02833	0.04220	0.04763	0.04953	0.05153	0.04350	0.07617
11	$\Theta_1$	0.02767	0.04260	0.04763	0.04920	0.05133	0.04277	0.07154
12	$\Theta_1$	0.01800	0.03033	0.03563	0.04280	0.05007	0.03463	0.04601
13	$\Theta_1$	0.02260	0.03907	0.04750	0.04873	0.05100	0.03930	0.05855
14	$\Theta_1$	0.02213	0.03393	0.04677	0.04940	0.05100	0.03930	0.06149
15	$\Theta_1$	0.01847	0.01847	0.03630	0.05040	0.05040	0.03557	0.04898
16	$\Theta_1$	0.01967	0.03560	0.04483	0.04820	0.05067	0.03703	0.05410
17	$\Theta_1$	0.02180	0.03787	0.04483	0.04860	0.05087	0.03877	0.05731
18	$\Theta_1$	0.02007	0.02867	0.04157	0.04907	0.05060	0.03697	0.05203

19	$\Theta_1$	0.02180	0.03847	0.04757	0.04873	0.05100	0.03910	0.05909
20	$\Theta_1$	0.02027	0.03733	0.04523	0.04833	0.05080	0.03783	0.05517
21	$\Theta_1$	0.02273	0.03947	0.04750	0.04880	0.05107	0.03963	0.06019
22	$\Theta_1$	0.02553	0.04133	0.04757	0.04913	0.05120	0.04157	0.06653
23	$\Theta_1$	0.02280	0.03940	0.04750	0.04887	0.05107	0.03997	0.06446
24	$\Theta_1$	0.02047	0.03747	0.04523	0.04827	0.05080	0.03783	0.05431
25	$\Theta_1$	0.01813	0.02713	0.03343	0.04567	0.05020	0.03470	0.04634
26	$\Theta_1$	0.02267	0.03840	0.04757	0.04920	0.05113	0.03990	0.06180
27	$\Theta_1$	0.02480	0.03967	0.04763	0.04927	0.05127	0.04110	0.06548
28	$\Theta_1$	0.02953	0.04253	0.04770	0.04960	0.05147	0.04377	0.07729
29	$\Theta_1$	0.02300	0.03853	0.04750	0.04880	0.05100	0.03950	0.05932
30	$\Theta_1$	0.02120	0.02933	0.04450	0.04967	0.05080	0.03837	0.05601
31	$\Theta_1$	0.02620	0.04073	0.04770	0.04947	0.05133	0.04203	0.06953
32	$\Theta_1$	0.02513	0.04067	0.04763	0.04913	0.05120	0.04097	0.06538
33	$\Theta_1$	0.02733	0.04127	0.04763	0.04940	0.05133	0.04263	0.07101
34	$\Theta_1$	0.02193	0.03880	0.04750	0.04860	0.05093	0.03897	0.05772
35	$\Theta_1$	0.01900	0.03700	0.04570	0.04827	0.05080	0.03723	0.05504
36	$\Theta_1$	0.01913	0.03447	0.04163	0.04793	0.05060	0.03657	0.05135
37	$\Theta_1$	0.02400	0.03933	0.04757	0.04933	0.05120	0.04070	0.06528
38	$\Theta_1$	0.01567	0.03133	0.04003	0.04767	0.05053	0.03477	0.05153
39	$\Theta_1$	0.01520	0.02067	0.02830	0.03860	0.04887	0.03083	0.03847
40	$\Theta_1$	0.02333	0.03960	0.04757	0.04933	0.05127	0.04097	0.06826
41	$\Theta_1$	0.02827	0.04193	0.04777	0.04960	0.05140	0.04317	0.07247

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.01567	0.02540	0.03090	0.03627	0.04933	0.03190	0.04064
43	$\Theta_1$	0.02507	0.04040	0.04763	0.04940	0.05133	0.04177	0.06949
44	$\Theta_1$	0.01853	0.03400	0.04203	0.04733	0.05047	0.03603	0.04988
45	$\Theta_1$	0.01860	0.03460	0.04063	0.04767	0.05067	0.03630	0.05184
46	$\Theta_1$	0.01613	0.02600	0.03370	0.04140	0.04987	0.03317	0.04454
47	$\Theta_1$	0.02453	0.03980	0.04750	0.04893	0.05107	0.04050	0.06111
48	$\Theta_1$	0.02047	0.03700	0.04590	0.04847	0.05073	0.03790	0.05607
49	$\Theta_1$	0.02553	0.03660	0.04763	0.05000	0.05127	0.04163	0.06743
50	$\Theta_1$	0.02327	0.04027	0.04757	0.04907	0.05113	0.04043	0.06586
51	$\Theta_1$	0.01840	0.03333	0.03937	0.04507	0.05053	0.03590	0.04978
52	$\Theta_1$	0.01900	0.03433	0.03990	0.04793	0.05060	0.03650	0.05175
53	$\Theta_1$	0.02587	0.04067	0.04763	0.04940	0.05133	0.04203	0.06934
54	$\Theta_1$	0.02700	0.04253	0.04757	0.04920	0.05140	0.04270	0.07455
55	$\Theta_1$	0.01807	0.03073	0.03863	0.04313	0.05020	0.03510	0.04739
56	$\Theta_1$	0.01867	0.03227	0.03817	0.04653	0.05040	0.03577	0.04944
57	$\Theta_1$	0.01627	0.02880	0.03717	0.04293	0.05040	0.03437	0.04711
58	$\Theta_1$	0.02160	0.03700	0.04390	0.04860	0.05087	0.03857	0.05700
59	$\Theta_1$	0.02933	0.04220	0.04777	0.04953	0.05147	0.04350	0.07472
60	$\Theta_1$	0.02413	0.03940	0.04770	0.04933	0.05127	0.04077	0.06643
61	$\Theta_1$	0.02960	0.04247	0.04770	0.04960	0.05147	0.04370	0.07495

62	$\Theta_1$	0.02247	0.03900	0.04703	0.04860	0.05093	0.03917	0.05819
63	$\Theta_1$	0.02700	0.04127	0.04770	0.04960	0.05140	0.04250	0.07337
64	$\Theta_1$	0.01820	0.02900	0.03543	0.04520	0.05020	0.03497	0.04725
65	$\Theta_1$	0.02647	0.04060	0.04770	0.04940	0.05133	0.04197	0.06742
66	$\Theta_1$	0.01667	0.02653	0.03177	0.03813	0.04980	0.03303	0.04299
67	$\Theta_1$	0.02533	0.04127	0.04763	0.04920	0.05127	0.04143	0.06810
68	$\Theta_1$	0.01580	0.02893	0.03950	0.04387	0.05047	0.03437	0.04841
69	$\Theta_1$	0.02193	0.03773	0.04737	0.04873	0.05093	0.03890	0.05810
70	$\Theta_1$	0.02353	0.03913	0.04757	0.04900	0.05107	0.04023	0.06210
71	$\Theta_1$	0.02413	0.03880	0.04757	0.04933	0.05120	0.04070	0.06396
72	$\Theta_1$	0.01780	0.03027	0.03530	0.04520	0.05033	0.03503	0.04764
73	$\Theta_1$	0.02293	0.03827	0.04757	0.04900	0.05100	0.03970	0.06106
74	$\Theta_1$	0.02340	0.03847	0.04757	0.04900	0.05113	0.04003	0.06067
75	$\Theta_1$	0.02533	0.04027	0.04763	0.04940	0.05133	0.04163	0.06775
76	$\Theta_1$	0.01907	0.01907	0.04023	0.05047	0.05047	0.03603	0.04940
77	$\Theta_1$	0.01780	0.02913	0.03757	0.04520	0.05013	0.03477	0.04663
78	$\Theta_1$	0.01787	0.03373	0.04077	0.04760	0.05060	0.03590	0.05058
79	$\Theta_1$	0.03260	0.04387	0.04783	0.04973	0.05153	0.04510	0.08390
80	$\Theta_1$	0.02873	0.04320	0.04763	0.04940	0.05147	0.04337	0.07331
81	$\Theta_1$	0.02027	0.03700	0.04010	0.04807	0.05073	0.03737	0.05356
82	$\Theta_1$	0.02540	0.04000	0.04763	0.04933	0.05127	0.04143	0.06616
83	$\Theta_1$	0.01767	0.03033	0.03543	0.04560	0.05027	0.03490	0.04812
84	$\Theta_1$	0.03013	0.04267	0.04777	0.04960	0.05140	0.04397	0.07565

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.02147	0.03653	0.04457	0.04853	0.05080	0.03830	0.05545
86	$\Theta_1$	0.02507	0.04027	0.04763	0.04940	0.05133	0.04163	0.07033
87	$\Theta_1$	0.01747	0.02773	0.03397	0.04127	0.05000	0.03410	0.04503
88	$\Theta_1$	0.02113	0.03833	0.04750	0.04873	0.05087	0.03850	0.05846
89	$\Theta_1$	0.02547	0.04007	0.04763	0.04933	0.05127	0.04150	0.06717
90	$\Theta_1$	0.02407	0.03967	0.04763	0.04933	0.05133	0.04103	0.06701
91	$\Theta_1$	0.02520	0.04107	0.04757	0.04900	0.05120	0.04130	0.06448
92	$\Theta_1$	0.02193	0.03793	0.04743	0.04887	0.05093	0.03910	0.05895
93	$\Theta_1$	0.01533	0.02207	0.02903	0.04467	0.04993	0.03263	0.04231
94	$\Theta_1$	0.02447	0.04007	0.04750	0.04907	0.05120	0.04077	0.06321
95	$\Theta_1$	0.02673	0.04187	0.04770	0.04953	0.05140	0.04263	0.07366
96	$\Theta_1$	0.01533	0.02480	0.02977	0.03767	0.04967	0.03203	0.04108
97	$\Theta_1$	0.02613	0.04080	0.04763	0.04940	0.05140	0.04217	0.07095
98	$\Theta_1$	0.02047	0.03813	0.04757	0.04873	0.05100	0.03830	0.05916
99	$\Theta_1$	0.02047	0.03693	0.04603	0.04847	0.05080	0.03790	0.05535
100	$\Theta_1$	0.02400	0.03973	0.04757	0.04940	0.05120	0.04110	0.06817
All	$\Theta_1$	0.04580	0.04773	0.04883	0.04993	0.05173	0.04890	0.05393

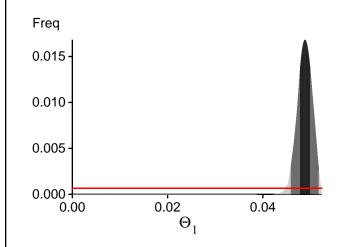
#### Citation suggestions:

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	-14972.22	-14538.78	-14566.18	-14627.95
2	-14095.10	-13931.51	-13989.28	-14064.98
3	-14952.03	-14718.03	-14779.06	-14852.22
4	-14332.99	-14124.26	-14188.18	-14252.02
5	-14497.83	-14263.53	-14317.47	-14391.54
6	-14181.39	-13991.90	-14055.62	-14121.05
7	-14061.56	-13902.57	-13963.74	-14037.57
8	-14418.53	-14160.45	-14213.96	-14278.85
9	-14637.68	-14333.03	-14379.02	-14445.10
10	-39594.78	-28448.21	-26586.32	-26638.53
11	-16126.56	-15374.19	-15355.61	-15413.11
12	-14338.60	-14137.16	-14196.37	-14266.85
13	-14378.57	-14126.84	-14181.63	-14247.95
14	-14590.05	-14338.63	-14393.45	-14459.43
15	-14240.97	-14017.73	-14070.20	-14141.43
16	-15706.92	-15068.44	-15057.78	-15126.30
17	-14571.11	-14313.21	-14369.06	-14435.35
18	-14157.36	-13961.73	-14022.18	-14090.33
19	-14846.58	-14474.25	-14510.12	-14574.77
20	-14546.91	-14243.74	-14286.89	-14356.77
21	-14252.85	-14064.20	-14129.71	-14194.77
22	-14608.98	-14285.99	-14332.53	-14393.64
23	-22832.27	-21183.35	-21055.77	-21116.48
24	-14741.50	-14378.07	-14411.47	-14479.05
25	-14237.82	-14003.08	-14052.38	-14124.11
26	-14369.30	-14144.82	-14204.23	-14270.03
27	-17087.67	-15786.19	-15662.24	-15723.12
28	-15906.78	-15316.82	-15331.86	-15385.96
29	-15221.47	-14596.58	-14585.66	-14651.56

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 14:46:07]

30	-14393.98	-14193.02	-14255.50	-14322.78
31	-14983.48	-14636.96	-14685.98	-14745.05
32	-15516.41	-14808.61	-14785.46	-14848.14
33	-20133.07	-18713.40	-18604.64	-18663.50
34	-14549.19	-14256.28	-14304.40	-14369.32
35	-14273.73	-14075.71	-14135.37	-14207.41
36	-14341.40	-14105.48	-14157.83	-14228.25
37	-15462.84	-15118.75	-15170.26	-15232.23
38	-21732.65	-20588.82	-20531.00	-20602.12
39	-14197.50	-13986.63	-14039.52	-14113.11
40	-18042.48	-16857.08	-16765.58	-16828.37
41	-14948.88	-14562.64	-14010.41	-14661.19
42	-14131.73	-13963.07	-13963.07	-14096.84
43	-20513.93	-19184.08	-14070.91	-19159.08
44	-14645.53	-14298.52	-14073.26	-14402.51
45	-14503.93	-14242.21	-14023.18	-14360.83
46	-14036.88	-13874.28	-13934.73	-14007.28
47	-14600.08	-14301.40	-14062.25	-14414.15
48	-14193.61	-14021.82	-14088.47	-14154.94
49	-15770.24	-15351.58	-14209.53	-15454.23
50	-16849.44	-15686.05	-14589.79	-15645.26
51	-14133.37	-13952.29	-14014.24	-14086.75
52	-14286.59	-14056.70	-14111.44	-14179.18
53	-14426.69	-14198.15	-14193.56	-14323.04
54	-18159.57	-17036.12	-14218.22	-17022.44
55	-14398.95	-14152.55	-14044.21	-14275.27
56	-14180.07	-13971.66	-14026.65	-14097.19
57	-14062.35	-13901.27	-13962.06	-14034.41
58	-14317.01	-14109.93	-14170.90	-14238.03
59	-14942.78	-14558.91	-14174.11	-14657.37
60	-16097.56	-15546.53	-14334.27	-15625.41
61	-17617.47	-16373.58	-14380.92	-16329.89
62	-14278.02	-14066.89	-14128.63	-14193.98
63	-14719.50	-14472.37	-14203.08	-14596.95
64	-14128.01	-13941.36	-14001.35	-14071.50
65	-14991.90	-14547.60	-14313.76	-14635.65
66	-14176.58	-14001.80	-14060.59	-14133.41
67	-15721.09	-15103.23	-14403.04	-15162.00
68	-15149.53	-14748.18	-14559.37	-14846.99
69	-16949.17	-15560.57	-14261.10	-15478.62
70	-14619.70	-14280.88	-14321.14	-14386.07
71	-14557.86	-14288.36	-14343.90	-14407.90
72	-14077.17	-13908.22	-13969.71	-14042.16
73	-14295.61	-14088.53	-14153.21	-14216.73
74	-15073.06	-14654.68	-14684.03	-14747.28

75	-26808.88	-21548.16	-15170.39	-20789.28
76	-14572.96	-14279.61	-14323.90	-14393.42
77	-14203.23	-13981.81	-14033.75	-14106.08
78	-14162.47	-13974.18	-14034.40	-14104.71
79	-81249.30	-51869.44	-21078.75	-46824.16
80	-14740.47	-14487.53	-13953.58	-14610.63
81	-16664.78	-15707.96	-15644.49	-15710.66
82	-15094.26	-14692.98	-14035.35	-14789.83
83	-14484.54	-14199.56	-13963.01	-14314.38
84	-15227.17	-14792.54	-14102.63	-14889.10
85	-14334.53	-14102.88	-14017.67	-14225.96
86	-15103.99	-14719.30	-14042.63	-14819.23
87	-14237.73	-14003.54	-14054.38	-14125.89
88	-14343.94	-14132.25	-14193.67	-14259.32
89	-14479.43	-14224.06	-14283.50	-14344.28
90	-14604.66	-14343.37	-14355.94	-14463.93
91	-14814.08	-14404.29	-14435.52	-14496.59
92	-15850.35	-14945.66	-14008.65	-14948.57
93	-14174.84	-13963.57	-14016.27	-14088.89
94	-16001.03	-15303.38	-14131.10	-15351.19
95	-14989.46	-14662.19	-14269.87	-14772.76
96	-14137.27	-13966.67	-14026.54	-14100.35
97	-14564.67	-14311.64	-14182.68	-14432.78
98	-14447.97	-14241.72	-13975.66	-14369.63
99	-14556.32	-14264.81	-14065.25	-14378.86
100	-14465.04	-14240.40	-14304.58	-14368.30
All	-1613972.14	-1527909.30	-1467418.54	-1529516.18

- (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 = 28.219571]

#### 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
$\Theta_1$	383095469/400015675	0.95770
Genealogies	172752940/1599984325	0.10797

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.67412	5056453.24
Genealogies	0.17165	18938126.99

## Average temperatures during the run

#### 

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

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