## **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 05:39:36 2017

Program finished at Sun Aug 13 07:07:19 2017 [Runtime:0000:01:27:43]



### **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) 1814728351

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.1.0 Haplotyping is turned on: NO

Output file: outfile\_1.0\_0.7

Posterior distribution raw histogram file: bayesfile

Raw data from the MCMC run: bayesallfile\_1.0\_0.7

Print data: No

Print genealogies [only some for some data type]:

## Data summary

Data file: infile.1.0
Datatype: Sequence data
Number of loci: 100

Mutation		Markey	M. to Constant Laboratory
Locus S	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]
5	1	Jukes-Cantor	[Basefreq: =0.25]
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32	1	Jukes-Cantor	[Basefreq: =0.25]

[Basefreq: =0.25]

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Jukes-Cantor

Jukes-Cantor

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36	1 1	Jukes-Cantor	[Basefreq: =0.25] [Basefreq: =0.25]
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99	1	Jukes-Cantor	[Basefreq: =0.25]	
100	1	Jukes-Cantor	[Basefreq: =0.25]	
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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	
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	
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	
39	1	1	1.000	1.000	1.000	
40	1	1	1.000	1.000	1.000	
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	
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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	
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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	
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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	
Population		ı	1.000	1.000	Locus	Gene copies
1 Roman					1	10
1 Koman	3110111_0				2	10
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Total of all populations	1	10	
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	99	10
	100	10
Minuster F. O. On J. Mary Johnson on Fay, and J. Engagger van on OF 20:201		

# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.02627	0.04067	0.04763	0.04933	0.05133	0.04203	0.06689
2	$\Theta_1$	0.02733	0.04220	0.04763	0.04920	0.05127	0.04237	0.06705
3	$\Theta_1$	0.02960	0.04320	0.04770	0.04933	0.05140	0.04337	0.07067
4	$\Theta_1$	0.02867	0.04293	0.04770	0.04927	0.05140	0.04310	0.07015
5	$\Theta_1$	0.02873	0.04293	0.04763	0.04907	0.05133	0.04310	0.07201
6	$\Theta_1$	0.03007	0.03680	0.04770	0.05060	0.05147	0.04377	0.07368
7	$\Theta_1$	0.03127	0.04300	0.04777	0.04960	0.05147	0.04430	0.07573
8	$\Theta_1$	0.02267	0.03793	0.04750	0.04900	0.05100	0.03950	0.05881
9	$\Theta_1$	0.02253	0.03840	0.04657	0.04853	0.05093	0.03910	0.05636
10	$\Theta_1$	0.03313	0.04407	0.04790	0.04973	0.05160	0.04530	0.08217
11	$\Theta_1$	0.03133	0.04353	0.04777	0.04967	0.05153	0.04470	0.07905
12	$\Theta_1$	0.02540	0.03960	0.04757	0.04913	0.05113	0.04103	0.06357
13	$\Theta_1$	0.02960	0.04240	0.04777	0.04960	0.05147	0.04370	0.07438
14	$\Theta_1$	0.03347	0.04533	0.04797	0.04953	0.05173	0.04550	0.08327
15	$\Theta_1$	0.02533	0.03940	0.04757	0.04913	0.05113	0.04090	0.06342
16	$\Theta_1$	0.02913	0.04287	0.04777	0.04953	0.05140	0.04350	0.07409
17	$\Theta_1$	0.02887	0.04193	0.04770	0.04953	0.05140	0.04323	0.07246
18	$\Theta_1$	0.02840	0.04193	0.04770	0.04953	0.05140	0.04323	0.07400

19	$\Theta_1$	0.02200	0.03767	0.04750	0.04873	0.05100	0.03903	0.05864
20	$\Theta_1$	0.03027	0.04267	0.04777	0.04967	0.05147	0.04390	0.07321
21	$\Theta_1$	0.02720	0.04127	0.04763	0.04940	0.05140	0.04263	0.07045
22	$\Theta_1$	0.02700	0.04147	0.04770	0.04960	0.05140	0.04270	0.07145
23	$\Theta_1$	0.02900	0.04200	0.04770	0.04953	0.05147	0.04330	0.07203
24	$\Theta_1$	0.02247	0.03827	0.04710	0.04827	0.05080	0.03857	0.05440
25	$\Theta_1$	0.03047	0.04267	0.04777	0.04960	0.05147	0.04390	0.07321
26	$\Theta_1$	0.02780	0.04160	0.04770	0.04953	0.05140	0.04283	0.07020
27	$\Theta_1$	0.01800	0.03080	0.03710	0.04333	0.05033	0.03510	0.04677
28	$\Theta_1$	0.02940	0.04247	0.04777	0.04967	0.05153	0.04370	0.07503
29	$\Theta_1$	0.02640	0.04047	0.04763	0.04933	0.05127	0.04183	0.06625
30	$\Theta_1$	0.02607	0.04120	0.04757	0.04907	0.05113	0.04157	0.06493
31	$\Theta_1$	0.02747	0.04107	0.04770	0.04933	0.05140	0.04243	0.06873
32	$\Theta_1$	0.02933	0.04233	0.04770	0.04960	0.05147	0.04357	0.07201
33	$\Theta_1$	0.02840	0.04220	0.04777	0.04967	0.05153	0.04343	0.07683
34	$\Theta_1$	0.02760	0.04207	0.04763	0.04913	0.05127	0.04230	0.06846
35	$\Theta_1$	0.03287	0.04407	0.04790	0.04973	0.05153	0.04530	0.08127
36	$\Theta_1$	0.03127	0.04307	0.04777	0.04953	0.05153	0.04437	0.07686
37	$\Theta_1$	0.02820	0.04260	0.04770	0.04920	0.05133	0.04277	0.06976
38	$\Theta_1$	0.02347	0.03953	0.04750	0.04873	0.05100	0.03977	0.05869
39	$\Theta_1$	0.03020	0.04287	0.04770	0.04967	0.05147	0.04410	0.07621
40	$\Theta_1$	0.02820	0.04167	0.04763	0.04947	0.05140	0.04297	0.06939
41	$\Theta_1$	0.02527	0.04000	0.04757	0.04927	0.05127	0.04143	0.06467

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03127	0.04420	0.04770	0.04933	0.05153	0.04443	0.07634
43	$\Theta_1$	0.03220	0.04353	0.04777	0.04967	0.05160	0.04477	0.08022
44	$\Theta_1$	0.02447	0.03987	0.04757	0.04920	0.05113	0.04090	0.06335
45	$\Theta_1$	0.03113	0.04313	0.04783	0.04967	0.05153	0.04437	0.07481
46	$\Theta_1$	0.03140	0.04293	0.04777	0.04960	0.05147	0.04423	0.07651
47	$\Theta_1$	0.02800	0.04160	0.04763	0.04953	0.05133	0.04290	0.07100
48	$\Theta_1$	0.02540	0.04147	0.04763	0.04907	0.05133	0.04163	0.06856
49	$\Theta_1$	0.03007	0.04273	0.04777	0.04967	0.05153	0.04397	0.07624
50	$\Theta_1$	0.01933	0.03207	0.03837	0.04540	0.05047	0.03610	0.04846
51	$\Theta_1$	0.02887	0.04220	0.04770	0.04940	0.05133	0.04303	0.07022
52	$\Theta_1$	0.02733	0.04140	0.04770	0.04953	0.05140	0.04270	0.07091
53	$\Theta_1$	0.02700	0.04073	0.04763	0.04933	0.05133	0.04217	0.06742
54	$\Theta_1$	0.02673	0.04193	0.04757	0.04913	0.05127	0.04217	0.06683
55	$\Theta_1$	0.02827	0.04187	0.04770	0.04960	0.05147	0.04317	0.07175
56	$\Theta_1$	0.02860	0.04140	0.04770	0.04947	0.05133	0.04277	0.06888
57	$\Theta_1$	0.02833	0.04200	0.04763	0.04953	0.05147	0.04330	0.07281
58	$\Theta_1$	0.03073	0.04287	0.04783	0.04967	0.05153	0.04410	0.07685
59	$\Theta_1$	0.02193	0.03647	0.04290	0.04840	0.05073	0.03830	0.05374
60	$\Theta_1$	0.02967	0.04227	0.04777	0.04953	0.05147	0.04357	0.07272
61	$\Theta_1$	0.02260	0.04233	0.04770	0.04960	0.05207	0.04357	0.07274

62	$\Theta_1$	0.03020	0.04267	0.04783	0.04973	0.05147	0.04390	0.07465
63	$\Theta_1$	0.02580	0.03980	0.04757	0.04907	0.05113	0.04103	0.06143
64	$\Theta_1$	0.03087	0.04287	0.04777	0.04960	0.05140	0.04417	0.07557
65	$\Theta_1$	0.02933	0.04180	0.04770	0.04947	0.05140	0.04317	0.07062
66	$\Theta_1$	0.03053	0.04293	0.04783	0.04973	0.05153	0.04417	0.07767
67	$\Theta_1$	0.02987	0.04233	0.04777	0.04960	0.05147	0.04357	0.07273
68	$\Theta_1$	0.02907	0.04173	0.04763	0.04940	0.05133	0.04310	0.07004
69	$\Theta_1$	0.02647	0.04100	0.04757	0.04913	0.05113	0.04163	0.06420
70	$\Theta_1$	0.02700	0.04173	0.04757	0.04900	0.05127	0.04190	0.06486
71	$\Theta_1$	0.02627	0.04047	0.04757	0.04933	0.05127	0.04183	0.06725
72	$\Theta_1$	0.02880	0.04187	0.04770	0.04953	0.05140	0.04323	0.07093
73	$\Theta_1$	0.02640	0.04147	0.04757	0.04900	0.05127	0.04163	0.06457
74	$\Theta_1$	0.03127	0.04387	0.04777	0.04940	0.05147	0.04417	0.07486
75	$\Theta_1$	0.03073	0.04393	0.04777	0.04940	0.05147	0.04417	0.07504
76	$\Theta_1$	0.02427	0.03887	0.04757	0.04920	0.05113	0.04037	0.06118
77	$\Theta_1$	0.02333	0.03873	0.04757	0.04913	0.05113	0.04023	0.06211
78	$\Theta_1$	0.02933	0.04300	0.04770	0.04927	0.05147	0.04330	0.07182
79	$\Theta_1$	0.02720	0.04107	0.04763	0.04940	0.05133	0.04243	0.06943
80	$\Theta_1$	0.03220	0.04467	0.04783	0.04953	0.05153	0.04483	0.08052
81	$\Theta_1$	0.02853	0.04153	0.04770	0.04947	0.05133	0.04290	0.06895
82	$\Theta_1$	0.02120	0.03667	0.04517	0.04827	0.05067	0.03783	0.05417
83	$\Theta_1$	0.03020	0.04287	0.04777	0.04967	0.05153	0.04410	0.07653
84	$\Theta_1$	0.02887	0.04200	0.04770	0.04960	0.05147	0.04330	0.07206

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.02673	0.04060	0.04757	0.04933	0.05127	0.04203	0.06661
86	$\Theta_1$	0.02940	0.04253	0.04777	0.04960	0.05153	0.04377	0.07325
87	$\Theta_1$	0.02520	0.04087	0.04757	0.04920	0.05127	0.04143	0.06642
88	$\Theta_1$	0.02807	0.04287	0.04770	0.04947	0.05147	0.04317	0.07435
89	$\Theta_1$	0.03273	0.04380	0.04783	0.04973	0.05167	0.04503	0.07907
90	$\Theta_1$	0.02793	0.04153	0.04763	0.04940	0.05140	0.04290	0.07053
91	$\Theta_1$	0.02980	0.04253	0.04770	0.04967	0.05147	0.04377	0.07348
92	$\Theta_1$	0.02687	0.04093	0.04770	0.04947	0.05140	0.04223	0.06785
93	$\Theta_1$	0.02960	0.04333	0.04770	0.04927	0.05140	0.04350	0.07467
94	$\Theta_1$	0.03200	0.04353	0.04783	0.04967	0.05160	0.04477	0.07949
95	$\Theta_1$	0.03133	0.04340	0.04777	0.04967	0.05153	0.04463	0.07729
96	$\Theta_1$	0.03320	0.04393	0.04790	0.04980	0.05160	0.04510	0.07931
97	$\Theta_1$	0.03167	0.04333	0.04777	0.04967	0.05147	0.04450	0.07766
98	$\Theta_1$	0.02527	0.03947	0.04757	0.04907	0.05113	0.04097	0.06166
99	$\Theta_1$	0.03047	0.04287	0.04777	0.04967	0.05153	0.04410	0.07809
100	$\Theta_1$	0.02467	0.03940	0.04757	0.04913	0.05120	0.04090	0.06279
All	$\Theta_1$	0.04773	0.04893	0.04990	0.05080	0.05200	0.04997	0.07702

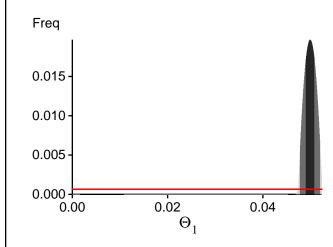
Citation suggestions:

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	-15588.83	-15149.64	-15194.78	-15244.47
2	-16758.65	-15892.42	-15866.23	-15915.16
3	-16730.96	-16014.94	-16020.72	-16067.70
4	-16432.08	-15963.79	-16016.57	-16063.97
5	-15970.71	-15472.90	-15514.46	-15563.10
6	-16028.86	-15489.68	-15523.76	-15578.62
7	-16692.23	-15967.58	-15974.09	-16018.31
8	-16433.76	-15855.58	-15879.48	-15932.23
9	-15693.77	-15038.21	-15037.54	-15090.60
10	-18689.12	-17367.93	-17280.39	-17326.49
11	-16109.92	-15695.06	-15759.99	-15801.52
12	-16056.41	-15451.25	-15467.76	-15517.80
13	-16405.20	-15717.13	-15726.41	-15771.14
14	-17155.00	-16595.98	-16647.36	-16684.88
15	-15973.81	-15450.83	-15484.56	-15535.04
16	-15769.36	-15325.31	-15376.79	-15421.72
17	-16493.29	-15865.55	-15887.44	-15933.69
18	-15654.79	-15310.41	-15380.65	-15426.97
19	-15965.04	-15382.24	-15400.43	-15452.28
20	-17161.37	-16093.52	-16033.50	-16078.71
21	-16305.44	-15607.09	-15610.36	-15659.75
22	-15872.51	-15404.22	-15450.87	-15497.15
23	-15668.29	-15204.86	-15250.51	-15299.61
24	-16078.31	-15308.79	-15287.79	-15341.90
25	-17342.19	-16209.83	-16139.05	-16184.82
26	-16128.77	-15598.25	-15634.29	-15681.24
27	-15122.26	-14714.86	-14754.84	-14812.00
28	-19563.50	-17595.37	-17380.25	-17425.21
29	-15714.48	-15202.90	-15236.23	-15284.76

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 05:39:36]

30	-15076.26	-14809.51	-14884.02	-14935.02
31	-16385.09	-15598.60	-15583.98	-15633.41
32	-15843.94	-15429.59	-15487.01	-15532.94
33	-16086.29	-15688.81	-15755.00	-15798.47
34	-16702.60	-15828.07	-15800.29	-15848.56
35	-18641.14	-17470.51	-17411.49	-17452.27
36	-17238.15	-16245.99	-16204.44	-16249.34
37	-15689.11	-15200.23	-15239.52	-15286.79
38	-15427.10	-14950.28	-14983.83	-15036.19
39	-17270.34	-16407.40	-16391.97	-16436.81
40	-16891.41	-15957.42	-15919.38	-15967.86
41	-15538.36	-15072.43	-15112.21	-15162.85
42	-16767.83	-16049.83	-16057.69	-16104.55
43	-19180.56	-17456.52	-17290.42	-17331.10
44	-14998.72	-14691.77	-14756.86	-14807.01
45	-17221.59	-16323.24	-16299.91	-16343.97
46	-16354.77	-15725.10	-15746.95	-15789.70
47	-16241.74	-15586.35	-15598.62	-15646.56
48	-15614.34	-15265.28	-15331.98	-15379.04
49	-17548.84	-16584.73	-16552.84	-16596.21
50	-15243.18	-14928.37	-14989.67	-15044.77
51	-15662.25	-15218.12	-15266.63	-15313.02
52	-16141.78	-15581.01	-15611.82	-15659.18
53	-17269.66	-16405.69	-16385.45	-16435.19
54	-16607.44	-15907.11	-15912.51	-15962.38
55	-16480.27	-15811.98	-15824.44	-15871.49
56	-17398.93	-16144.60	-16046.32	-16094.16
57	-17271.76	-16420.02	-16403.38	-16451.54
58	-16912.10	-16178.58	-16186.43	-16233.84
59	-16267.09	-15621.35	-15629.19	-15682.41
60	-16082.04	-15518.96	-15549.62	-15594.93
61	-17199.78	-16183.34	-16133.62	-16179.55
62	-16172.18	-15572.45	-15596.38	-15641.98
63	-16884.95	-15859.65	-15799.25	-15852.85
64	-18468.78	-17236.75	-17159.81	-17204.95
65	-15779.78	-15301.43	-15343.87	-15391.86
66	-16637.38	-16030.25	-16060.38	-16106.41
67	-17265.69	-16515.59	-16520.98	-16568.04
68	-15838.75	-15425.54	-15481.59	-15529.43
69	-15972.83	-15415.26	-15442.88	-15491.92
70	-15724.49	-15156.20	-15178.55	-15231.24
71	-15707.76	-15278.82	-15330.03	-15378.13
72	-17590.77	-16464.77	-16395.04	-16443.23
73	-15796.35	-15242.48	-15267.58	-15318.02
74	-16825.94	-16006.23	-15992.42	-16038.23

All	-1656773.64	-1584326.49	-1584510.69	-1589257.73
100	-15307.77	-14918.16	-14969.27	-15021.30
99	-17614.60	-16861.81	-16875.85	-16916.35
98	-15376.33	-14933.39	-14975.85	-15026.07
97	-22673.57	-19163.24	-18671.44	-18715.06
96	-16909.70	-16116.75	-16113.43	-16155.09
95	-17573.27	-16592.69	-16558.25	-16601.76
94	-16976.82	-16235.54	-16244.87	-16285.57
93	-15884.64	-15530.06	-15602.05	-15645.48
92	-15592.07	-15202.80	-15260.01	-15309.01
91	-16123.65	-15607.01	-15646.78	-15693.31
90	-15792.89	-15342.57	-15390.40	-15438.43
89	-18384.71	-17233.87	-17173.75	-17217.13
88	-16469.79	-15902.91	-15938.30	-15983.85
87	-15663.59	-15277.41	-15335.46	-15384.13
86	-17286.02	-16299.33	-16257.41	-16302.97
85	-16711.74	-15964.79	-15962.12	-16012.26
84	-20624.78	-18046.87	-17718.26	-17765.06
83	-16525.75	-15953.07	-15987.73	-16033.88
82	-16054.28	-15457.34	-15472.97	-15526.27
81	-15722.89	-15282.12	-15331.29	-15379.00
80	-18368.27	-17519.61	-17523.05	-17564.96
79	-16163.68	-15575.26	-15597.90	-15646.04
78	-16049.56	-15459.76	-15483.78	-15529.83
77	-15394.98	-15005.10	-15056.60	-15109.39
76	-15108.38	-14766.54	-14824.88	-14876.83
75	-16985.93	-16411.26	-16450.73	-16495.51

- (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 = 45.054771]

#### 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	386397354/399985555 72645146/1600014445	0.96603 0.04540

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.61436	2397742.70
Genealogies	0.27260	5732072.53

## Average temperatures during the run

#### 

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

4

0.00000

### 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