# **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 09:55:31 2017

Program finished at Sun Aug 13 11:14:16 2017 [Runtime:0000:01:18:45]



## **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) 2423382887

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

**Exponential Distribution** -Population size estimation:

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 Delta Prior Minimum Mean Maximum Bins UpdateFreq Theta -11 Uniform 0.000000 0.050 0.100 0.010 1500 0.20000

[-1 -1 means priors were set globally]

Number of concurrent chains (replicates) [c]

Markov chain settings: Long chain

Number of chains 50000 Recorded steps [a] 200 Increment (record every x step [b]

20000000 Visited (sampled) parameter values [a\*b\*c] 10000 Number of discard trees per chain (burn-in)

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: NO

Output file: outfile\_0.8\_0.4

Posterior distribution raw histogram file: bayesfile

bayesallfile\_0.8\_0.4 Print data: No

Print genealogies [only some for some data type]: None

Raw data from the MCMC run:

# Data summary

Data file:

Datatype:

Sequence data

Number of loci:

100

Mutationmodel:

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]	
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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	
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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	
27	1	1	1.000	1.000	1.000	
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46	1	1	1.000	1.000	1.000	
47	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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77	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	
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91	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	
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97	1	1	1.000	1.000	1.000	
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# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.03440	0.04460	0.04797	0.05000	0.05167	0.04570	0.08644
2	$\Theta_1$	0.03213	0.04413	0.04790	0.04993	0.05167	0.04523	0.08422
3	$\Theta_1$	0.03480	0.04467	0.04797	0.04993	0.05153	0.04577	0.08663
4	$\Theta_1$	0.03307	0.04400	0.04790	0.04980	0.05167	0.04517	0.08432
5	$\Theta_1$	0.03247	0.04433	0.04803	0.05000	0.05173	0.04550	0.08648
6	$\Theta_1$	0.03407	0.04480	0.04777	0.04960	0.05160	0.04597	0.08699
7	$\Theta_1$	0.03400	0.04427	0.04770	0.04960	0.05153	0.04543	0.08579
8	$\Theta_1$	0.03393	0.04473	0.04797	0.04987	0.05167	0.04583	0.08586
9	$\Theta_1$	0.03313	0.04393	0.04810	0.04987	0.05173	0.04510	0.08454
10	$\Theta_1$	0.03353	0.04433	0.04790	0.04987	0.05160	0.04550	0.08447
11	$\Theta_1$	0.03280	0.04360	0.04770	0.04933	0.05160	0.04497	0.08634
12	$\Theta_1$	0.03353	0.04447	0.04797	0.04973	0.05167	0.04570	0.08620
13	$\Theta_1$	0.03300	0.04453	0.04797	0.04993	0.05167	0.04570	0.08533
14	$\Theta_1$	0.03173	0.04447	0.04790	0.04973	0.05193	0.04570	0.08714
15	$\Theta_1$	0.03300	0.04420	0.04777	0.04967	0.05160	0.04543	0.08556
16	$\Theta_1$	0.03393	0.04500	0.04810	0.05020	0.05160	0.04603	0.08683
17	$\Theta_1$	0.03293	0.04467	0.04790	0.04993	0.05173	0.04577	0.08506
18	$\Theta_1$	0.03493	0.04467	0.04797	0.04973	0.05173	0.04590	0.08685

19	$\Theta_1$	0.03533	0.04447	0.04777	0.04967	0.05147	0.04563	0.08781
20	$\Theta_1$	0.03553	0.04567	0.04790	0.04927	0.05153	0.04583	0.08734
21	$\Theta_1$	0.03353	0.04447	0.04777	0.04967	0.05167	0.04563	0.08546
22	$\Theta_1$	0.03260	0.04420	0.04783	0.04980	0.05167	0.04537	0.08391
23	$\Theta_1$	0.03567	0.04513	0.04803	0.05000	0.05173	0.04617	0.08752
24	$\Theta_1$	0.03360	0.04560	0.04783	0.04927	0.05173	0.04583	0.08687
25	$\Theta_1$	0.03433	0.04513	0.04817	0.05007	0.05173	0.04617	0.08811
26	$\Theta_1$	0.03340	0.04460	0.04790	0.04980	0.05160	0.04577	0.08732
27	$\Theta_1$	0.03360	0.04447	0.04783	0.04980	0.05173	0.04563	0.08631
28	$\Theta_1$	0.03280	0.04387	0.04770	0.04973	0.05160	0.04503	0.08490
29	$\Theta_1$	0.03493	0.04507	0.04803	0.05000	0.05160	0.04610	0.08711
30	$\Theta_1$	0.03300	0.04360	0.04790	0.04960	0.05167	0.04490	0.08448
31	$\Theta_1$	0.03433	0.04553	0.04790	0.04933	0.05167	0.04570	0.08730
32	$\Theta_1$	0.03293	0.04453	0.04803	0.05000	0.05160	0.04557	0.08539
33	$\Theta_1$	0.03507	0.04553	0.04790	0.04940	0.05147	0.04570	0.08809
34	$\Theta_1$	0.03267	0.04407	0.04783	0.04980	0.05160	0.04523	0.08576
35	$\Theta_1$	0.03187	0.04387	0.04790	0.04973	0.05167	0.04503	0.08343
36	$\Theta_1$	0.03240	0.04413	0.04790	0.04987	0.05173	0.04530	0.08463
37	$\Theta_1$	0.03347	0.04453	0.04797	0.04993	0.05167	0.04563	0.08590
38	$\Theta_1$	0.03273	0.04433	0.04790	0.04987	0.05153	0.04550	0.08456
39	$\Theta_1$	0.03227	0.04413	0.04803	0.05000	0.05167	0.04523	0.08550
40	$\Theta_1$	0.03300	0.04420	0.04783	0.04973	0.05153	0.04530	0.08416
41	$\Theta_1$	0.03187	0.04360	0.04790	0.04980	0.05167	0.04483	0.08425

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03400	0.04480	0.04810	0.04993	0.05180	0.04597	0.08705
43	$\Theta_1$	0.03393	0.04493	0.04797	0.04993	0.05153	0.04597	0.08715
44	$\Theta_1$	0.03247	0.04400	0.04790	0.04987	0.05167	0.04517	0.08539
45	$\Theta_1$	0.03347	0.04473	0.04797	0.04987	0.05167	0.04590	0.08645
46	$\Theta_1$	0.03420	0.04473	0.04803	0.05000	0.05173	0.04583	0.08700
47	$\Theta_1$	0.03320	0.04460	0.04797	0.04993	0.05160	0.04570	0.08639
48	$\Theta_1$	0.03107	0.04387	0.04783	0.04993	0.05160	0.04497	0.08309
49	$\Theta_1$	0.03400	0.04433	0.04783	0.04973	0.05153	0.04543	0.08645
50	$\Theta_1$	0.03387	0.04473	0.04797	0.04993	0.05167	0.04577	0.08510
51	$\Theta_1$	0.03440	0.04533	0.04810	0.05027	0.05187	0.04630	0.08878
52	$\Theta_1$	0.03133	0.04453	0.04790	0.04973	0.05180	0.04570	0.08776
53	$\Theta_1$	0.03413	0.04433	0.04790	0.04967	0.05153	0.04557	0.08658
54	$\Theta_1$	0.03347	0.04547	0.04797	0.04967	0.05160	0.04570	0.08665
55	$\Theta_1$	0.03373	0.04440	0.04790	0.04980	0.05153	0.04557	0.08539
56	$\Theta_1$	0.03273	0.04413	0.04797	0.04987	0.05160	0.04530	0.08465
57	$\Theta_1$	0.03373	0.04433	0.04790	0.04987	0.05167	0.04543	0.08561
58	$\Theta_1$	0.03273	0.04440	0.04810	0.05000	0.05167	0.04550	0.08591
59	$\Theta_1$	0.03013	0.04307	0.04777	0.04953	0.05147	0.04397	0.07903
60	$\Theta_1$	0.03280	0.04400	0.04797	0.04993	0.05160	0.04510	0.08571
61	$\Theta_1$	0.03453	0.04600	0.04810	0.04967	0.05180	0.04617	0.08857

62	$\Theta_1$	0.03513	0.04547	0.04810	0.04960	0.05153	0.04577	0.08709
63	$\Theta_1$	0.03340	0.04480	0.04803	0.04993	0.05167	0.04597	0.08655
64	$\Theta_1$	0.03267	0.04373	0.04790	0.04967	0.05160	0.04503	0.08231
65	$\Theta_1$	0.03467	0.04480	0.04790	0.04973	0.05180	0.04603	0.08659
66	$\Theta_1$	0.03220	0.04400	0.04790	0.04980	0.05167	0.04517	0.08433
67	$\Theta_1$	0.03280	0.04460	0.04810	0.05007	0.05173	0.04563	0.08659
68	$\Theta_1$	0.03460	0.04447	0.04797	0.04987	0.05187	0.04563	0.08691
69	$\Theta_1$	0.03380	0.04427	0.04803	0.04993	0.05160	0.04543	0.08577
70	$\Theta_1$	0.03233	0.04400	0.04790	0.04980	0.05160	0.04517	0.08409
71	$\Theta_1$	0.03527	0.04487	0.04790	0.04973	0.05160	0.04603	0.08783
72	$\Theta_1$	0.03313	0.04447	0.04810	0.05007	0.05167	0.04550	0.08485
73	$\Theta_1$	0.03300	0.04513	0.04797	0.04960	0.05160	0.04530	0.08528
74	$\Theta_1$	0.03427	0.04487	0.04797	0.04967	0.05180	0.04610	0.08787
75	$\Theta_1$	0.03433	0.04573	0.04803	0.04960	0.05180	0.04590	0.08756
76	$\Theta_1$	0.03220	0.04500	0.04783	0.04953	0.05167	0.04517	0.08391
77	$\Theta_1$	0.03540	0.04480	0.04803	0.04993	0.05160	0.04590	0.08846
78	$\Theta_1$	0.03460	0.04507	0.04803	0.04993	0.05153	0.04610	0.08659
79	$\Theta_1$	0.03327	0.04533	0.04783	0.04940	0.05167	0.04557	0.08502
80	$\Theta_1$	0.03360	0.04467	0.04803	0.04993	0.05180	0.04577	0.08629
81	$\Theta_1$	0.03333	0.04440	0.04803	0.04993	0.05167	0.04550	0.08508
82	$\Theta_1$	0.03313	0.04400	0.04797	0.04973	0.05160	0.04523	0.08552
83	$\Theta_1$	0.03253	0.04420	0.04790	0.04987	0.05173	0.04530	0.08630
84	$\Theta_1$	0.03407	0.04440	0.04790	0.04967	0.05160	0.04563	0.08625

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.03467	0.04460	0.04797	0.04980	0.05153	0.04577	0.08676
86	$\Theta_1$	0.03387	0.04440	0.04790	0.04987	0.05167	0.04557	0.08579
87	$\Theta_1$	0.03480	0.04547	0.04803	0.04980	0.05160	0.04563	0.08738
88	$\Theta_1$	0.03347	0.04420	0.04790	0.04980	0.05167	0.04537	0.08548
89	$\Theta_1$	0.03413	0.04487	0.04803	0.05000	0.05187	0.04597	0.08654
90	$\Theta_1$	0.03253	0.04380	0.04777	0.04973	0.05153	0.04497	0.08473
91	$\Theta_1$	0.03520	0.04553	0.04817	0.05027	0.05187	0.04643	0.08750
92	$\Theta_1$	0.03347	0.04453	0.04797	0.04987	0.05153	0.04563	0.08707
93	$\Theta_1$	0.03307	0.04453	0.04790	0.04980	0.05167	0.04570	0.08622
94	$\Theta_1$	0.03220	0.04380	0.04777	0.04973	0.05153	0.04497	0.08387
95	$\Theta_1$	0.03253	0.04393	0.04783	0.04973	0.05160	0.04510	0.08365
96	$\Theta_1$	0.03280	0.04447	0.04790	0.04993	0.05160	0.04557	0.08503
97	$\Theta_1$	0.03527	0.04453	0.04803	0.04980	0.05167	0.04570	0.08657
98	$\Theta_1$	0.03287	0.04413	0.04797	0.04980	0.05173	0.04530	0.08646
99	$\Theta_1$	0.03347	0.04433	0.04790	0.04993	0.05160	0.04550	0.08624
100	$\Theta_1$	0.03287	0.04467	0.04790	0.04993	0.05173	0.04577	0.08592
All	$\Theta_1$	0.01167	0.01573	0.01823	0.01953	0.02180	0.01737	0.09977

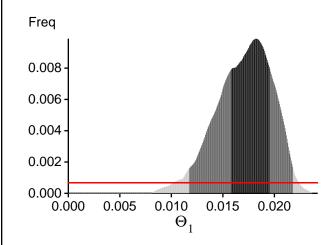
#### Citation suggestions:

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	-15259.90	-14907.17	-14942.81	-15012.86
2	-14188.67	-13978.51	-14021.72	-14101.89
3	-14579.14	-14249.14	-14274.73	-14350.70
4	-15582.59	-15121.26	-15134.85	-15213.06
5	-14687.43	-14362.71	-14403.26	-14472.26
6	-15405.19	-14817.37	-14813.52	-14881.02
7	-14617.78	-14296.81	-14332.77	-14406.09
8	-14285.65	-14097.91	-14149.55	-14228.34
9	-14132.82	-13948.65	-14000.55	-14078.40
10	-14395.22	-14108.30	-14143.26	-14221.45
11	-16440.12	-15733.67	-15700.53	-15771.65
12	-14823.38	-14538.86	-14583.96	-14656.18
13	-14746.00	-14436.11	-14469.98	-14544.58
14	-14625.22	-14376.41	-14427.39	-14498.57
15	-15240.85	-14810.34	-14833.40	-14902.16
16	-16417.38	-15312.29	-15206.47	-15279.73
17	-14976.83	-14698.88	-14729.20	-14812.41
18	-14383.37	-14152.67	-14204.41	-14277.17
19	-16385.39	-15497.57	-15429.56	-15502.73
20	-14695.57	-14369.52	-14408.85	-14476.35
21	-16365.34	-15509.31	-15461.21	-15529.61
22	-14081.32	-13906.04	-13954.87	-14037.64
23	-15035.99	-14535.49	-14545.31	-14610.42
24	-14514.41	-14293.91	-14341.64	-14413.98
25	-15411.66	-14761.55	-14746.74	-14813.55
26	-15499.62	-14929.82	-14925.67	-14997.92
27	-15509.59	-14791.91	-14748.94	-14825.29
28	-14568.83	-14258.26	-14294.90	-14370.37
29	-37698.30	-24661.88	-22361.31	-22436.34

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 09:55:31]

30	-15703.42	-14777.21	-14697.19	-14775.09
31	-14624.02	-14314.68	-14352.99	-14424.48
32	-14500.54	-14195.86	-14220.87	-14302.74
33	-19572.43	-18155.66	-18029.21	-18094.16
34	-21259.18	-18879.59	-18564.51	-18636.52
35	-14063.27	-13905.52	-13955.64	-14039.13
36	-14451.36	-14142.54	-14174.00	-14251.28
37	-16886.68	-16042.81	-15993.75	-16068.32
38	-14272.58	-14044.35	-14089.31	-14168.45
39	-15210.88	-14805.50	-14825.81	-14897.60
40	-14106.19	-13929.84	-13978.27	-14062.26
41	-15215.23	-14733.71	-14740.51	-14817.76
42	-15076.71	-14657.70	-14678.35	-14746.13
43	-15279.78	-14809.32	-14818.91	-14887.51
44	-14323.16	-14110.97	-14164.09	-14237.43
45	-15254.61	-14753.78	-14766.22	-14833.53
46	-16000.86	-15085.21	-15014.85	-15083.45
47	-15065.02	-14491.43	-14478.84	-14551.63
48	-14117.63	-13963.64	-14006.76	-14094.80
49	-14412.33	-14200.85	-14252.74	-14326.17
50	-14590.75	-14342.71	-14388.15	-14465.72
51	-52248.18	-32445.26	-28952.55	-29010.87
52	-23300.82	-19673.99	-19147.34	-19206.55
53	-14657.93	-14293.70	-14317.10	-14390.72
54	-14351.03	-14139.98	-14194.31	-14264.94
55	-28133.73	-23921.99	-23322.79	-23394.24
56	-15454.01	-14651.55	-14593.65	-14671.14
57	-14532.86	-14298.39	-14332.93	-14414.77
58	-15193.06	-14760.42	-14774.87	-14848.79
59	-13931.88	-13774.20	-13818.87	-13908.43
60	-14317.28	-14113.24	-14155.55	-14235.42
61	-15852.58	-14983.87	-14930.26	-14993.48
62	-14653.20	-14338.06	-14367.07	-14443.88
63	-14639.36	-14328.28	-14360.55	-14435.69
64	-14075.59	-13899.35	-13947.73	-14033.11
65	-14385.12	-14129.00	-14174.53	-14248.47
66	-14121.46	-13931.43	-13979.47	-14059.69
67	-14726.29	-14388.05	-14423.13	-14492.35
68	-14858.60	-14462.81	-14478.52	-14556.60
69	-14710.60	-14373.77	-14403.90	-14477.47
70	-14100.04	-13924.84	-13977.76	-14056.75
71	-15872.70	-15079.55	-15041.93	-15104.41
72	-14351.94	-14060.34	-14093.83	-14171.49
73	-14508.91	-14247.65	-14286.64	-14362.45
74	-15117.80	-14623.21	-14636.68	-14701.59

All	-1597739.13	-1509467.88	-1502790.73	-1510189.48
100	-20094.43	-17489.47	-17127.36	-17195.24
99	-14513.31	-14206.84	-14233.13	-14316.61
98	-14675.70	-14346.66	-14371.16	-14452.33
97	-14819.30	-14462.42	-14494.55	-14565.14
96	-14610.92	-14243.64	-14266.39	-14342.24
95	-14830.23	-14449.64	-14471.25	-14548.82
94	-14631.82	-14240.27	-14242.91	-14332.54
93	-27360.81	-22935.81	-22247.19	-22330.32
92	-14983.64	-14594.38	-14618.62	-14687.75
91	-15338.08	-14752.66	-14752.93	-14815.44
90	-14194.51	-14006.01	-14058.58	-14135.98
89	-14557.26	-14277.42	-14319.20	-14390.64
88	-14361.47	-14169.42	-14221.66	-14299.68
87	-15692.02	-15205.05	-15221.44	-15290.44
86	-14788.75	-14341.72	-14350.01	-14424.73
85	-14522.19	-14198.05	-14235.57	-14303.51
84	-15109.47	-14787.58	-14821.76	-14895.41
83	-17249.69	-15989.77	-15867.51	-15935.77
82	-14315.73	-14099.96	-14145.52	-14220.80
81	-14190.73	-14003.11	-14052.40	-14131.83
80	-14555.69	-14241.00	-14279.06	-14348.69
79	-14320.64	-14044.37	-14082.07	-14158.21
78	-15720.60	-14872.86	-14814.25	-14884.70
77	-15086.78	-14739.24	-14785.23	-14845.80
76	-14061.21	-13903.79	-13956.71	-14037.19
75	-14678.81	-14378.48	-14419.85	-14489.34

- (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 = 159.837824]

#### 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
$\Theta_1$	371948628/400030693	0.92980
Genealogies	179297091/1599969307	0.11206

# MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.44801	3820798.79
Genealogies	0.45251	3975727.46

# Average temperatures during the run

# Chain Temperatures 1 0.00000 2 0.00000 3 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

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