# **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 21:40:47 2017

Program finished at Sun Aug 13 23:14:44 2017 [Runtime:0000:01:33:57]



### **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) 1584829672

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

Print options:

Data file: infile.1.0
Haplotyping is turned on: NO

Output file: outfile\_1.0\_0.4

Posterior distribution raw histogram file: bayesfile

Raw data from the MCMC run: bayesallfile\_1.0\_0.4

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

Mutationmodel:

Mutation	model:			
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]	
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3	1 1	1.000	1.000	1.000	
4	1 1	1.000	1.000	1.000	
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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	
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# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.03607	0.04493	0.04810	0.04993	0.05187	0.04610	0.08914
2	$\Theta_1$	0.03713	0.04527	0.04810	0.04980	0.05147	0.04637	0.08961
3	$\Theta_1$	0.03067	0.04533	0.04803	0.04967	0.05187	0.04557	0.08718
4	$\Theta_1$	0.03373	0.04460	0.04783	0.04973	0.05167	0.04583	0.08862
5	$\Theta_1$	0.03460	0.04533	0.04817	0.05020	0.05167	0.04637	0.08875
6	$\Theta_1$	0.03213	0.04447	0.04797	0.04993	0.05167	0.04550	0.08787
7	$\Theta_1$	0.03467	0.04433	0.04790	0.05007	0.05167	0.04543	0.08765
8	$\Theta_1$	0.03047	0.04487	0.04783	0.04980	0.05200	0.04603	0.08873
9	$\Theta_1$	0.03620	0.04527	0.04777	0.04953	0.05180	0.04657	0.08911
10	$\Theta_1$	0.03573	0.04520	0.04783	0.04967	0.05173	0.04630	0.08852
11	$\Theta_1$	0.03540	0.04567	0.04790	0.04933	0.05153	0.04583	0.08759
12	$\Theta_1$	0.03527	0.04533	0.04837	0.05033	0.05173	0.04630	0.08868
13	$\Theta_1$	0.03647	0.04520	0.04810	0.05000	0.05180	0.04630	0.08893
14	$\Theta_1$	0.03440	0.04527	0.04797	0.04987	0.05180	0.04637	0.08901
15	$\Theta_1$	0.03400	0.04540	0.04783	0.04933	0.05160	0.04570	0.08789
16	$\Theta_1$	0.03480	0.04473	0.04783	0.04987	0.05140	0.04577	0.08811
17	$\Theta_1$	0.03673	0.04500	0.04803	0.04987	0.05160	0.04617	0.08854
18	$\Theta_1$	0.03520	0.04580	0.04783	0.04927	0.05173	0.04603	0.08835

19	$\Theta_1$	0.03647	0.04593	0.04823	0.05020	0.05167	0.04690	0.08887
20	$\Theta_1$	0.03593	0.04533	0.04790	0.04993	0.05160	0.04637	0.08827
21	$\Theta_1$	0.03380	0.04540	0.04810	0.05013	0.05193	0.04643	0.08872
22	$\Theta_1$	0.03280	0.04447	0.04797	0.05007	0.05167	0.04550	0.08761
23	$\Theta_1$	0.03633	0.04660	0.04823	0.04960	0.05173	0.04683	0.08891
24	$\Theta_1$	0.03813	0.04513	0.04783	0.04947	0.05160	0.04643	0.08914
25	$\Theta_1$	0.03480	0.04487	0.04803	0.04993	0.05167	0.04597	0.08807
26	$\Theta_1$	0.03567	0.04500	0.04803	0.04987	0.05167	0.04610	0.08882
27	$\Theta_1$	0.03427	0.04507	0.04817	0.05007	0.05180	0.04610	0.08830
28	$\Theta_1$	0.03533	0.04553	0.04817	0.05013	0.05180	0.04663	0.08921
29	$\Theta_1$	0.03547	0.04587	0.04803	0.04947	0.05173	0.04603	0.08894
30	$\Theta_1$	0.03687	0.04553	0.04810	0.05000	0.05180	0.04657	0.08922
31	$\Theta_1$	0.03347	0.04453	0.04810	0.05000	0.05180	0.04563	0.08693
32	$\Theta_1$	0.03627	0.04553	0.04830	0.05007	0.05193	0.04657	0.08938
33	$\Theta_1$	0.03487	0.04607	0.04817	0.04960	0.05187	0.04643	0.08808
34	$\Theta_1$	0.03540	0.04487	0.04790	0.04973	0.05167	0.04597	0.08858
35	$\Theta_1$	0.03533	0.04527	0.04803	0.05007	0.05173	0.04630	0.08888
36	$\Theta_1$	0.03527	0.04500	0.04810	0.05007	0.05167	0.04603	0.08842
37	$\Theta_1$	0.03413	0.04427	0.04790	0.04967	0.05160	0.04550	0.08758
38	$\Theta_1$	0.03527	0.04513	0.04817	0.04993	0.05167	0.04623	0.08855
39	$\Theta_1$	0.03393	0.04473	0.04790	0.04987	0.05167	0.04577	0.08823
40	$\Theta_1$	0.03580	0.04527	0.04797	0.04980	0.05167	0.04643	0.08910
41	$\Theta_1$	0.03633	0.04613	0.04797	0.04913	0.05167	0.04630	0.08824

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03360	0.04473	0.04777	0.04973	0.05167	0.04583	0.08752
43	$\Theta_1$	0.03460	0.04487	0.04817	0.05007	0.05167	0.04597	0.08732
44	$\Theta_1$	0.03380	0.04420	0.04797	0.04973	0.05173	0.04543	0.08727
45	$\Theta_1$	0.03320	0.04513	0.04797	0.04953	0.05173	0.04550	0.08718
46	$\Theta_1$	0.03387	0.04420	0.04817	0.04980	0.05173	0.04543	0.08877
47	$\Theta_1$	0.03460	0.04527	0.04797	0.04953	0.05160	0.04557	0.08824
48	$\Theta_1$	0.03247	0.04420	0.04777	0.04960	0.05173	0.04543	0.08778
49	$\Theta_1$	0.03507	0.04520	0.04797	0.04993	0.05180	0.04623	0.08862
50	$\Theta_1$	0.03493	0.04513	0.04810	0.05013	0.05180	0.04610	0.08849
51	$\Theta_1$	0.03667	0.04607	0.04790	0.04933	0.05160	0.04637	0.08914
52	$\Theta_1$	0.03453	0.04473	0.04803	0.05007	0.05167	0.04577	0.08701
53	$\Theta_1$	0.03553	0.04487	0.04797	0.04953	0.05180	0.04623	0.08899
54	$\Theta_1$	0.03280	0.04593	0.04797	0.04947	0.05173	0.04610	0.08799
55	$\Theta_1$	0.03660	0.04587	0.04797	0.04933	0.05167	0.04630	0.08923
56	$\Theta_1$	0.03560	0.04467	0.04810	0.04993	0.05167	0.04583	0.08809
57	$\Theta_1$	0.03500	0.04433	0.04783	0.04927	0.05173	0.04577	0.08952
58	$\Theta_1$	0.03487	0.04533	0.04803	0.05013	0.05167	0.04630	0.08877
59	$\Theta_1$	0.03453	0.04480	0.04810	0.04993	0.05180	0.04590	0.08867
60	$\Theta_1$	0.03640	0.04507	0.04810	0.04967	0.05180	0.04637	0.08864
61	$\Theta_1$	0.03533	0.04513	0.04797	0.04993	0.05167	0.04617	0.08793

62	$\Theta_1$	0.03500	0.04440	0.04783	0.04947	0.05173	0.04577	0.08739
63	$\Theta_1$	0.03380	0.04507	0.04803	0.04987	0.05167	0.04617	0.08775
64	$\Theta_1$	0.02827	0.04440	0.04797	0.04967	0.05220	0.04563	0.08849
65	$\Theta_1$	0.03673	0.04493	0.04803	0.04973	0.05167	0.04617	0.08863
66	$\Theta_1$	0.03433	0.04440	0.04790	0.04973	0.05147	0.04563	0.08716
67	$\Theta_1$	0.03400	0.04513	0.04803	0.04993	0.05167	0.04617	0.08853
68	$\Theta_1$	0.03793	0.04573	0.04803	0.04993	0.05173	0.04670	0.08931
69	$\Theta_1$	0.03247	0.04447	0.04817	0.05000	0.05173	0.04557	0.08733
70	$\Theta_1$	0.03680	0.04513	0.04783	0.04980	0.05160	0.04623	0.08823
71	$\Theta_1$	0.03533	0.04467	0.04783	0.04973	0.05187	0.04590	0.08790
72	$\Theta_1$	0.03367	0.04467	0.04797	0.04980	0.05167	0.04583	0.08774
73	$\Theta_1$	0.03467	0.04493	0.04803	0.05000	0.05167	0.04597	0.08755
74	$\Theta_1$	0.03433	0.04473	0.04777	0.04967	0.05167	0.04590	0.08803
75	$\Theta_1$	0.03400	0.04473	0.04797	0.04993	0.05153	0.04590	0.08836
76	$\Theta_1$	0.03600	0.04513	0.04803	0.04987	0.05173	0.04630	0.08892
77	$\Theta_1$	0.03773	0.04527	0.04777	0.04953	0.05173	0.04650	0.08932
78	$\Theta_1$	0.03587	0.04547	0.04817	0.05013	0.05167	0.04643	0.08895
79	$\Theta_1$	0.03620	0.04500	0.04790	0.04967	0.05180	0.04617	0.08884
80	$\Theta_1$	0.03600	0.04533	0.04837	0.05040	0.05160	0.04630	0.08864
81	$\Theta_1$	0.03507	0.04480	0.04797	0.04987	0.05167	0.04597	0.08779
82	$\Theta_1$	0.03627	0.04580	0.04783	0.04913	0.05153	0.04597	0.08869
83	$\Theta_1$	0.03520	0.04520	0.04797	0.04987	0.05167	0.04623	0.08805
84	$\Theta_1$	0.03660	0.04447	0.04763	0.04920	0.05160	0.04590	0.08923

85 86 87 88 89	Parameter $\Theta_{1}$ $\Theta_{1}$ $\Theta_{1}$ $\Theta_{1}$ $\Theta_{1}$ $\Theta_{1}$ $\Theta_{1}$	2.5% 0.03533 0.03467 0.03427 0.03513 0.03547	25.0% 0.04507 0.04453 0.04467 0.04513 0.04493	0.04817 0.04790 0.04803 0.04810	75.0% 0.05007 0.04987 0.04980 0.05013	97.5% 0.05167 0.05153 0.05173	0.04610 0.04563 0.04583	Mean 0.08875 0.08769 0.08902
86 87 88 89 90	$\Theta_1$ $\Theta_1$ $\Theta_1$ $\Theta_1$	0.03467 0.03427 0.03513	0.04453 0.04467 0.04513	0.04790	0.04987	0.05153	0.04563	0.08769
87 88 89 90	$\Theta_1$ $\Theta_1$ $\Theta_1$	0.03427	0.04467	0.04803	0.04980			
88 89 90	$\Theta_1$ $\Theta_1$	0.03513	0.04513			0.05173	0.04583	0.08902
89 90	$\Theta_1$			0.04810	0.05013			
90		0.03547	0.04493		0.03013	0.05173	0.04617	0.08883
	$\Theta_1$		0.0 / 100	0.04810	0.04993	0.05167	0.04610	0.08859
		0.03500	0.04467	0.04810	0.04980	0.05167	0.04583	0.08810
91	$\Theta_1$	0.03653	0.04513	0.04810	0.05007	0.05160	0.04617	0.08939
92	$\Theta_1$	0.03620	0.04540	0.04803	0.04973	0.05160	0.04657	0.08939
93	$\Theta_1$	0.03433	0.04520	0.04810	0.05007	0.05173	0.04623	0.08875
94	$\Theta_1$	0.03587	0.04473	0.04777	0.04960	0.05153	0.04603	0.08818
95	$\Theta_1$	0.03467	0.04427	0.04783	0.04947	0.05173	0.04563	0.08780
96	$\Theta_1$	0.03460	0.04453	0.04763	0.04947	0.05167	0.04583	0.08882
97	$\Theta_1$	0.03427	0.04573	0.04803	0.04947	0.05173	0.04597	0.08911
98	$\Theta_1$	0.03340	0.04460	0.04810	0.05027	0.05147	0.04557	0.08839
99	$\Theta_1$	0.03627	0.04473	0.04803	0.04953	0.05173	0.04603	0.08837
100	$\Theta_1$	0.03460	0.04527	0.04837	0.05027	0.05180	0.04630	0.08941
All	$\Theta_1$	0.01440	0.02047	0.02290	0.02420	0.02720	0.02203	0.09983

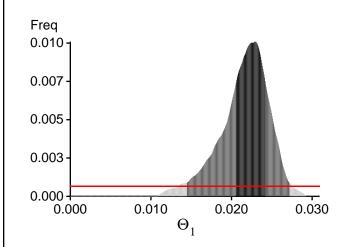
Citation suggestions:

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	-17436.33	-16244.22	-16153.75	-16200.31
2	-17989.45	-16680.15	-16563.84	-16624.53
3	-15758.70	-15273.17	-15294.24	-15355.15
4	-16012.70	-15313.89	-15299.68	-15359.36
5	-16760.97	-15778.70	-15711.89	-15773.85
6	-17728.50	-16373.26	-16245.69	-16304.72
7	-17252.23	-15994.49	-15877.90	-15940.05
8	-16515.68	-15840.52	-15830.76	-15895.93
9	-16233.43	-15429.61	-15396.95	-15453.13
10	-15937.96	-15449.44	-15462.05	-15526.17
11	-15929.37	-15234.18	-15212.71	-15276.02
12	-17212.71	-16043.89	-15946.09	-16000.83
13	-16660.26	-15889.16	-15874.55	-15924.66
14	-16603.97	-15583.87	-15516.46	-15568.38
15	-16813.87	-15870.77	-15815.90	-15872.03
16	-15861.00	-15360.04	-15387.88	-15443.99
17	-15815.53	-15091.62	-15066.25	-15125.57
18	-16975.79	-15854.19	-15753.57	-15818.26
19	-16759.53	-15913.25	-15868.60	-15933.38
20	-15562.94	-15141.43	-15167.78	-15230.71
21	-16990.76	-15854.73	-15768.71	-15823.32
22	-15900.43	-15373.86	-15398.61	-15450.63
23	-17809.94	-16556.80	-16454.73	-16506.39
24	-17225.75	-16314.98	-16283.62	-16329.76
25	-17914.59	-16558.30	-16426.43	-16490.39
26	-17443.79	-16437.91	-16382.28	-16438.03
27	-16217.97	-15611.45	-15615.00	-15678.55
28	-17493.11	-16299.16	-16211.49	-16260.54
29	-17476.96	-16362.95	-16290.45	-16341.89

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 21:40:47]

30	-16691.04	-15712.20	-15655.95	-15707.14
31	-15938.65	-15303.33	-15289.04	-15353.93
32	-15964.10	-15500.46	-15524.86	-15588.43
33	-15018.48	-14618.81	-14642.88	-14712.64
34	-15675.57	-15227.45	-15258.88	-15315.25
35	-15843.08	-15314.15	-15337.28	-15390.96
36	-15820.99	-15297.44	-15319.20	-15374.38
37	-16998.02	-16176.31	-16145.82	-16203.75
38	-15767.45	-15103.96	-15095.33	-15152.93
39	-16812.06	-15513.13	-15378.88	-15446.79
40	-16675.25	-15859.93	-15835.91	-15885.87
41	-15623.61	-15079.61	-15085.04	-15147.86
42	-16057.30	-15385.50	-15364.47	-15434.27
43	-15085.59	-14659.20	-14678.75	-14750.16
44	-15658.26	-15159.48	-15170.39	-15233.45
45	-18077.30	-16161.72	-15915.93	-15981.92
46	-17616.05	-16435.10	-16343.57	-16394.62
47	-16378.69	-15708.64	-15701.12	-15756.50
48	-15473.29	-15097.02	-15135.43	-15200.49
49	-16432.09	-15772.47	-15768.68	-15831.05
50	-15662.74	-15186.73	-15214.35	-15272.05
51	-16555.45	-15685.76	-15643.00	-15694.57
52	-15964.86	-15242.86	-15219.00	-15281.40
53	-18253.66	-16832.65	-16705.93	-16756.54
54	-16239.33	-15656.38	-15669.57	-15721.56
55	-17203.23	-16376.32	-16344.54	-16412.88
56	-16961.24	-15578.40	-15425.22	-15493.63
57	-18567.92	-16957.45	-16777.88	-16843.57
58	-16390.54	-15752.13	-15756.88	-15811.58
59	-17222.46	-15991.17	-15880.65	-15936.76
60	-16224.41	-15420.57	-15386.33	-15443.83
61	-15678.59	-15119.80	-15113.43	-15180.39
62	-15686.14	-15152.46	-15168.80	-15229.94
63	-16567.35	-15554.65	-15478.69	-15542.19
64	-16773.47	-15950.81	-15907.26	-15971.73
65	-15386.52	-14854.83	-14855.36	-14925.64
66	-16021.10	-15444.04	-15436.78	-15507.80
67	-16462.89	-15499.45	-15434.43	-15494.85
68	-17900.74	-16524.81	-16399.42	-16450.40
69	-15343.81	-14877.23	-14894.88	-14960.52
70	-15136.19	-14770.84	-14809.50	-14870.22
71	-16958.32	-15927.44	-15858.36	-15913.41
72	-16226.46	-15441.14	-15404.53	-15466.86
73	-15255.42	-14817.80	-14841.79	-14904.38
74	-17091.37	-16350.09	-16323.29	-16392.18
L				

All	-1664847.69	-1575910.39	-1571129.65	-1577010.69
100	-17943.44	-16536.75	-16399.15	-16460.68
99	-18005.54	-16324.27	-16135.29	-16189.35
98	-16402.04	-15664.76	-15651.95	-15701.19
97	-16684.52	-15973.69	-15973.59	-16020.94
96	-18310.84	-17048.24	-16965.60	-17004.47
95	-15295.77	-14847.69	-14869.55	-14930.26
94	-15027.76	-14648.50	-14684.55	-14744.17
93	-15872.42	-15265.36	-15263.73	-15330.21
92	-17796.76	-16734.61	-16676.62	-16725.63
91	-23082.55	-18971.86	-18338.03	-18393.73
90	-15890.34	-15287.95	-15290.19	-15353.63
89	-16674.28	-15963.91	-15960.72	-16010.93
88	-15795.38	-15360.91	-15393.84	-15449.71
87	-17769.31	-16460.12	-16347.42	-16397.56
86	-16789.06	-15955.50	-15928.19	-15981.05
85	-17124.61	-15986.97	-15878.30	-15945.93
84	-17121.52	-16036.00	-15964.66	-16018.12
83	-15757.98	-15014.83	-14978.21	-15045.73
82	-16313.65	-15612.63	-15609.44	-15660.88
81	-16222.94	-15418.16	-15372.70	-15444.08
80	-15654.86	-15299.78	-15334.52	-15397.55
79	-18742.17	-17062.04	-16880.39	-16937.45
78	-16705.81	-15745.39	-15680.98	-15737.89
77	-16899.44	-16094.95	-16080.50	-16126.34
76	-17679.28	-16608.97	-16538.65	-16599.55
75	-17835.08	-16389.79	-16242.66	-16300.78

- (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 = 184.915730]

#### 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
$\Theta_1$	364503620/399989579	0.91128
Genealogies	66702008/1600010421	0.04169

# MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.37847	4513224.48
Genealogies	0.66179	2068076.57

# 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