## **AUTO**

POPULATION SIZE, MIGRATION, DIVERGENCE, ASSIGNMENT, HISTORY

Bayesian inference using the structured coalescent

Migrate-n version 5.0.0a [May-20-2017]

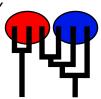
Using Intel AVX (Advanced Vector Extensions)

Compiled for PARALLEL computer architectures

One master and 100 compute nodes are available.

Program started at Sun Aug 13 04:05:37 2017

Program finished at Sun Aug 13 05:39:14 2017 [Runtime:0000:01:33:37]



### **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) 3120634400

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]

Markov chain settings: Long chain

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

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.0.9 NO

Haplotyping is turned on:

Output file: outfile\_0.9\_0.5

Posterior distribution raw histogram file: bayesfile Raw data from the MCMC run: bayesallfile\_0.9\_0.5

Print data: No

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

# Data summary

Data file: infile.0.9
Datatype: Sequence data
Number of loci: 100

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Locus Sublocus Mutationmodel Mutationmodel parameters  1    1    Jukes-Cantor [Basefreq: =0.25] 2    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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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	
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# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.03173	0.04407	0.04783	0.04980	0.05173	0.04523	0.08349
2	$\Theta_1$	0.03247	0.04433	0.04790	0.04987	0.05153	0.04543	0.08485
3	$\Theta_1$	0.03220	0.04393	0.04803	0.04987	0.05173	0.04517	0.08462
4	$\Theta_1$	0.03393	0.04420	0.04803	0.04993	0.05167	0.04537	0.08386
5	$\Theta_1$	0.03420	0.04473	0.04783	0.04973	0.05160	0.04583	0.08636
6	$\Theta_1$	0.02947	0.04287	0.04783	0.04980	0.05160	0.04403	0.07937
7	$\Theta_1$	0.03307	0.04440	0.04777	0.04953	0.05160	0.04563	0.08708
8	$\Theta_1$	0.03387	0.04453	0.04797	0.04987	0.05173	0.04570	0.08548
9	$\Theta_1$	0.03207	0.04393	0.04790	0.04967	0.05160	0.04470	0.08316
10	$\Theta_1$	0.03347	0.04533	0.04783	0.04940	0.05153	0.04550	0.08572
11	$\Theta_1$	0.03047	0.04313	0.04777	0.04967	0.05153	0.04430	0.08232
12	$\Theta_1$	0.03240	0.04420	0.04783	0.04973	0.05160	0.04537	0.08478
13	$\Theta_1$	0.03427	0.04433	0.04790	0.04980	0.05167	0.04550	0.08606
14	$\Theta_1$	0.03347	0.04420	0.04783	0.04973	0.05160	0.04537	0.08473
15	$\Theta_1$	0.03293	0.04520	0.04777	0.04927	0.05160	0.04537	0.08601
16	$\Theta_1$	0.03293	0.04427	0.04803	0.05000	0.05173	0.04537	0.08442
17	$\Theta_1$	0.03113	0.04393	0.04797	0.04987	0.05160	0.04510	0.08379
18	$\Theta_1$	0.03453	0.04460	0.04797	0.04980	0.05167	0.04577	0.08644

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 04:05:37]

19	$\Theta_1$	0.03327	0.04433	0.04790	0.04967	0.05167	0.04557	0.08673
20	$\Theta_1$	0.03300	0.04447	0.04803	0.05000	0.05180	0.04557	0.08501
21	$\Theta_1$	0.03113	0.04373	0.04783	0.04987	0.05160	0.04483	0.08157
22	$\Theta_1$	0.03173	0.04380	0.04797	0.04980	0.05173	0.04503	0.08237
23	$\Theta_1$	0.03467	0.04493	0.04783	0.04960	0.05167	0.04570	0.08649
24	$\Theta_1$	0.03487	0.04487	0.04790	0.04987	0.05160	0.04597	0.08691
25	$\Theta_1$	0.03413	0.04540	0.04797	0.04953	0.05160	0.04557	0.08443
26	$\Theta_1$	0.03420	0.04453	0.04783	0.04980	0.05160	0.04563	0.08745
27	$\Theta_1$	0.03187	0.04413	0.04783	0.04980	0.05153	0.04523	0.08355
28	$\Theta_1$	0.03320	0.04433	0.04790	0.04980	0.05173	0.04557	0.08433
29	$\Theta_1$	0.03447	0.04547	0.04783	0.04947	0.05167	0.04563	0.08604
30	$\Theta_1$	0.03160	0.04353	0.04783	0.04980	0.05160	0.04470	0.08249
31	$\Theta_1$	0.03167	0.04467	0.04777	0.04940	0.05153	0.04490	0.08388
32	$\Theta_1$	0.03220	0.04380	0.04790	0.04967	0.05173	0.04510	0.08351
33	$\Theta_1$	0.03320	0.04473	0.04803	0.05000	0.05160	0.04583	0.08587
34	$\Theta_1$	0.03240	0.04400	0.04790	0.04980	0.05167	0.04523	0.08377
35	$\Theta_1$	0.02900	0.04347	0.04790	0.04987	0.05180	0.04463	0.08163
36	$\Theta_1$	0.03173	0.04400	0.04803	0.05000	0.05160	0.04503	0.08345
37	$\Theta_1$	0.03253	0.04420	0.04790	0.04980	0.05167	0.04537	0.08435
38	$\Theta_1$	0.03433	0.04440	0.04777	0.04953	0.05153	0.04557	0.08503
39	$\Theta_1$	0.03320	0.04400	0.04783	0.04973	0.05167	0.04523	0.08410
40	$\Theta_1$	0.03387	0.04433	0.04777	0.04967	0.05153	0.04550	0.08458
41	$\Theta_1$	0.03167	0.04480	0.04790	0.04953	0.05153	0.04497	0.08297

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03140	0.04320	0.04777	0.04960	0.05160	0.04450	0.08185
43	$\Theta_1$	0.03433	0.04440	0.04777	0.04967	0.05160	0.04557	0.08654
44	$\Theta_1$	0.03273	0.04387	0.04783	0.04980	0.05153	0.04503	0.08334
45	$\Theta_1$	0.03020	0.04300	0.04777	0.04960	0.05147	0.04423	0.07892
46	$\Theta_1$	0.02953	0.04253	0.04777	0.04967	0.05153	0.04377	0.07911
47	$\Theta_1$	0.03380	0.04413	0.04797	0.04980	0.05167	0.04530	0.08679
48	$\Theta_1$	0.03447	0.04473	0.04777	0.04953	0.05160	0.04597	0.08679
49	$\Theta_1$	0.03273	0.04407	0.04803	0.04973	0.05167	0.04530	0.08538
50	$\Theta_1$	0.03400	0.04433	0.04783	0.04967	0.05167	0.04557	0.08618
51	$\Theta_1$	0.03407	0.04447	0.04797	0.04987	0.05160	0.04557	0.08542
52	$\Theta_1$	0.03187	0.04360	0.04777	0.04967	0.05160	0.04483	0.08256
53	$\Theta_1$	0.03300	0.04413	0.04790	0.04980	0.05160	0.04523	0.08288
54	$\Theta_1$	0.03367	0.04467	0.04797	0.04993	0.05187	0.04577	0.08607
55	$\Theta_1$	0.03367	0.04413	0.04803	0.04967	0.05180	0.04543	0.08609
56	$\Theta_1$	0.03360	0.04440	0.04790	0.04973	0.05160	0.04557	0.08449
57	$\Theta_1$	0.03080	0.04333	0.04790	0.04973	0.05167	0.04450	0.08257
58	$\Theta_1$	0.03267	0.04493	0.04790	0.04967	0.05160	0.04530	0.08413
59	$\Theta_1$	0.03173	0.04360	0.04790	0.04980	0.05160	0.04483	0.08270
60	$\Theta_1$	0.03287	0.04420	0.04790	0.04987	0.05160	0.04530	0.08333
61	$\Theta_1$	0.03160	0.04373	0.04783	0.04973	0.05160	0.04490	0.08308

62	$\Theta_1$	0.03387	0.04547	0.04803	0.04980	0.05173	0.04577	0.08548
63	$\Theta_1$	0.03127	0.04347	0.04777	0.04967	0.05160	0.04470	0.08311
64	$\Theta_1$	0.03453	0.04413	0.04790	0.04973	0.05160	0.04537	0.08703
65	$\Theta_1$	0.02753	0.04193	0.04770	0.04967	0.05147	0.04317	0.07475
66	$\Theta_1$	0.03267	0.04427	0.04783	0.04987	0.05167	0.04537	0.08428
67	$\Theta_1$	0.03153	0.04380	0.04790	0.04993	0.05167	0.04490	0.08200
68	$\Theta_1$	0.03167	0.04360	0.04790	0.04973	0.05153	0.04483	0.08424
69	$\Theta_1$	0.03400	0.04427	0.04783	0.04967	0.05160	0.04550	0.08604
70	$\Theta_1$	0.03173	0.04353	0.04790	0.04973	0.05160	0.04470	0.08148
71	$\Theta_1$	0.02993	0.04287	0.04783	0.04973	0.05153	0.04410	0.07893
72	$\Theta_1$	0.03293	0.04447	0.04797	0.04993	0.05173	0.04557	0.08557
73	$\Theta_1$	0.03160	0.04367	0.04783	0.04980	0.05167	0.04483	0.08435
74	$\Theta_1$	0.03173	0.04367	0.04777	0.04973	0.05153	0.04490	0.08219
75	$\Theta_1$	0.03320	0.04347	0.04777	0.04947	0.05147	0.04477	0.08445
76	$\Theta_1$	0.03280	0.03973	0.04803	0.05093	0.05167	0.04537	0.08430
77	$\Theta_1$	0.02980	0.04313	0.04783	0.04973	0.05153	0.04430	0.08042
78	$\Theta_1$	0.03260	0.04400	0.04777	0.04973	0.05160	0.04523	0.08446
79	$\Theta_1$	0.03493	0.04507	0.04803	0.05000	0.05167	0.04617	0.08730
80	$\Theta_1$	0.03167	0.04407	0.04783	0.04980	0.05167	0.04523	0.08366
81	$\Theta_1$	0.03047	0.04360	0.04790	0.04987	0.05160	0.04477	0.08153
82	$\Theta_1$	0.03033	0.04327	0.04790	0.04980	0.05167	0.04450	0.08107
83	$\Theta_1$	0.03213	0.04400	0.04777	0.04967	0.05160	0.04523	0.08472
84	$\Theta_1$	0.03433	0.04467	0.04790	0.04973	0.05160	0.04583	0.08721

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.02873	0.04260	0.04770	0.04973	0.05153	0.04377	0.07896
86	$\Theta_1$	0.03280	0.04493	0.04770	0.04927	0.05160	0.04517	0.08378
87	$\Theta_1$	0.03327	0.04420	0.04783	0.04967	0.05167	0.04543	0.08439
88	$\Theta_1$	0.03227	0.04347	0.04783	0.04973	0.05153	0.04470	0.08346
89	$\Theta_1$	0.03393	0.04420	0.04783	0.04967	0.05167	0.04543	0.08609
90	$\Theta_1$	0.03393	0.04513	0.04783	0.04933	0.05160	0.04537	0.08528
91	$\Theta_1$	0.03367	0.04427	0.04783	0.04967	0.05160	0.04550	0.08383
92	$\Theta_1$	0.03167	0.04340	0.04797	0.04980	0.05160	0.04463	0.08369
93	$\Theta_1$	0.03133	0.04427	0.04790	0.04993	0.05173	0.04537	0.08519
94	$\Theta_1$	0.03173	0.04340	0.04783	0.04967	0.05160	0.04463	0.08290
95	$\Theta_1$	0.03233	0.04413	0.04797	0.04987	0.05167	0.04530	0.08317
96	$\Theta_1$	0.02747	0.04293	0.04777	0.04940	0.05153	0.04310	0.07717
97	$\Theta_1$	0.03173	0.04393	0.04790	0.04993	0.05160	0.04503	0.08355
98	$\Theta_1$	0.03000	0.04307	0.04777	0.04967	0.05153	0.04430	0.08012
99	$\Theta_1$	0.03287	0.04487	0.04783	0.04960	0.05153	0.04510	0.08513
100	$\Theta_1$	0.03127	0.04227	0.04777	0.05007	0.05153	0.04463	0.08186
All	$\Theta_1$	0.00840	0.01307	0.01497	0.01673	0.01840	0.01430	0.09969

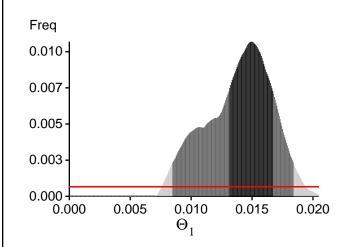
#### Citation suggestions:

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	-17384.85	-16097.41	-15980.72	-16041.72
2	-16448.29	-15865.08	-15885.20	-15941.93
3	-15829.21	-15165.48	-15151.85	-15216.53
4	-15501.84	-14946.34	-14954.12	-15019.36
5	-16541.55	-15471.14	-15392.36	-15448.35
6	-14318.72	-14091.03	-14146.84	-14214.83
7	-16270.00	-15506.00	-15489.21	-15539.23
8	-15156.88	-14731.77	-14767.27	-14823.42
9	-15170.13	-14820.97	-14869.95	-14929.67
10	-17212.78	-16253.45	-16207.11	-16263.50
11	-14929.50	-14605.39	-14653.93	-14715.04
12	-15420.31	-14916.16	-14939.12	-14996.09
13	-15600.78	-15110.91	-15138.70	-15193.86
14	-15140.48	-14771.08	-14815.09	-14874.31
15	-15923.16	-15314.45	-15325.41	-15379.04
16	-14881.53	-14520.46	-14562.04	-14620.71
17	-15436.29	-15059.53	-15106.13	-15165.02
18	-18341.91	-16589.05	-16397.66	-16450.29
19	-16613.31	-15849.99	-15839.03	-15890.28
20	-15523.31	-14949.32	-14957.31	-15016.43
21	-14943.31	-14604.97	-14645.50	-14710.92
22	-14661.15	-14314.37	-14354.27	-14416.99
23	-16157.71	-15379.08	-15360.17	-15412.04
24	-16923.63	-15954.31	-15905.17	-15957.45
25	-15325.81	-14844.21	-14865.52	-14927.75
26	-17366.26	-16189.95	-16103.59	-16155.09
27	-16266.15	-15220.98	-15138.50	-15200.33
28	-14772.76	-14524.37	-14587.89	-14647.22
29	-16126.39	-15463.38	-15467.56	-15518.21

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 04:05:37]

30	-14739.20	-14413.35	-14457.14	-14518.75
31	-15634.97	-15185.94	-15222.23	-15277.85
32	-14964.89	-14545.14	-14571.27	-14633.38
33	-17218.28	-16214.84	-16162.08	-16214.18
34	-15221.18	-14802.77	-14839.99	-14897.95
35	-15028.30	-14619.91	-14654.82	-14716.76
36	-15053.44	-14636.72	-14666.82	-14729.49
37	-14829.87	-14470.97	-14513.15	-14571.93
38	-15171.26	-14658.61	-14675.22	-14733.27
39	-15793.79	-15056.68	-15027.77	-15091.67
40	-17056.78	-15794.70	-15675.88	-15739.80
41	-14562.06	-14335.89	-14394.52	-14457.22
42	-14673.57	-14431.37	-14487.66	-14552.77
43	-18816.95	-17785.29	-17749.20	-17795.59
44	-15381.91	-14888.71	-14910.07	-14969.54
45	-14508.79	-14201.32	-14244.22	-14311.12
46	-15046.37	-14687.86	-14726.97	-14792.65
47	-26069.19	-20406.56	-19502.38	-19555.19
48	-17229.22	-16324.43	-16284.12	-16343.28
49	-20823.32	-17606.15	-17134.86	-17193.48
50	-16287.14	-15320.69	-15260.96	-15315.83
51	-15150.17	-14718.81	-14748.17	-14805.44
52	-15572.81	-15147.77	-15182.37	-15244.04
53	-15425.33	-14892.35	-14904.31	-14966.43
54	-15450.71	-15085.85	-15141.96	-15194.01
55	-16643.85	-15784.62	-15752.06	-15806.00
56	-15187.70	-14769.36	-14805.73	-14864.74
57	-15121.73	-14691.79	-14721.21	-14782.58
58	-14677.53	-14421.61	-14481.00	-14542.91
59	-14881.95	-14513.61	-14554.58	-14614.99
60	-15110.70	-14762.63	-14809.80	-14870.71
61	-14782.36	-14515.65	-14574.72	-14636.71
62	-16139.35	-15276.53	-15233.52	-15289.38
63	-15465.80	-14958.35	-14972.43	-15035.85
64	-19682.61	-18314.39	-18213.68	-18267.46
65	-14209.66	-13994.56	-14047.84	-14120.27
66	-16176.96	-15378.44	-15346.71	-15407.38
67	-14771.99	-14432.95	-14476.75	-14539.45
68	-15236.32	-14868.49	-14914.05	-14972.64
69	-35004.30	-23353.62	-21333.01	-21383.94
70	-15136.20	-14633.07	-14641.35	-14709.21
71	-14319.74	-14094.97	-14150.25	-14219.70
72	-16261.54	-15602.22	-15598.00	-15654.44
73	-15987.42	-15385.18	-15394.19	-15451.82
74	-15721.05	-15143.85	-15150.71	-15214.28

75	-14807.32	-14496.39	-14549.80	-14607.86
76	-16439.39	-15453.37	-15386.05	-15447.62
77	-14962.42	-14592.84	-14632.14	-14694.92
78	-15167.78	-14689.39	-14711.69	-14770.47
79	-16469.56	-15586.48	-15548.76	-15599.74
80	-15288.52	-14748.45	-14757.85	-14818.54
81	-14982.53	-14496.71	-14509.77	-14577.01
82	-15314.73	-14745.28	-14746.41	-14810.81
83	-15641.96	-15057.71	-15067.06	-15123.75
84	-15704.16	-15205.15	-15226.94	-15284.11
85	-14599.36	-14339.04	-14387.71	-14456.34
86	-19977.94	-17046.68	-16624.20	-16684.29
87	-15253.60	-14683.17	-14688.48	-14746.72
88	-18564.09	-17433.45	-17362.81	-17421.08
89	-15562.71	-15153.45	-15199.69	-15252.21
90	-15022.48	-14666.53	-14716.31	-14771.87
91	-15035.27	-14668.90	-14712.99	-14775.89
92	-15149.02	-14776.19	-14819.83	-14880.62
93	-15296.32	-14862.38	-14896.88	-14953.55
94	-15819.12	-15415.02	-15450.57	-15515.47
95	-15655.61	-15052.14	-15050.67	-15111.78
96	-15990.24	-15505.27	-15522.86	-15592.87
97	-14977.14	-14607.30	-14647.05	-14708.40
98	-14509.63	-14296.44	-14355.30	-14421.96
99	-16789.60	-15723.96	-15647.05	-15704.23
100	-21681.01	-20023.97	-19870.57	-19929.67
All	-1610936.27	-1531443.31	-1528264.89	-1534211.88

- (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 = 141.521205]

#### 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
$\Theta_1$	376232195/399972469	0.94065
Genealogies	107661963/1600027531	0.06729

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$ Genealogies	0.49133 0.37391	3423452.91 4634692.83

## 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
No warning was recorded during the run