# **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 Sun Jul 23 19:32:45 2017

Program finished at Sun Jul 23 22:37:02 2017 [Runtime:0000:03:04:17]



## **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) 1018040110

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:

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]:

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 19:32:45]

# Data summary

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

Mutationmodel:

Mutationmodel:				
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]	
6	1	Jukes-Cantor	[Basefreq: =0.25]	
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10	1	Jukes-Cantor	[Basefreq: =0.25]	
11	1	Jukes-Cantor	[Basefreq: =0.25]	
12	1	Jukes-Cantor	[Basefreq: =0.25]	
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80	1	Jukes-Cantor	[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	Sites
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	e variation and probab				
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	
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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	
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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	

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52	1	1	1.000	1.000	1.000	
53	1	1	1.000	1.000	1.000	
54	1	1	1.000	1.000	1.000	
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56	1	1	1.000	1.000	1.000	
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	
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	
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	
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
i Koman	3110111_0				2	10
					3	10
					4	10
					5	10
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	86	10	
	87	10	
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	89	10	
	90	10	
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	92	10	
	93	10	
	94	10	
	95	10	
	96	10	
	97	10	
	98	10	
	99	10	
	100	10	
Total of all populations	1	10	
	2	10	
	3	10	
	4	10	
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	6	10	
	7	10	
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95	10
96	10
97	10
98	10
99	10
100	10

# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.03253	0.04420	0.04797	0.04993	0.05180	0.04537	0.08713
2	$\Theta_1$	0.03460	0.04533	0.04823	0.05007	0.05187	0.04643	0.08977
3	$\Theta_1$	0.03540	0.04547	0.04790	0.04947	0.05167	0.04610	0.08785
4	$\Theta_1$	0.03533	0.04467	0.04777	0.04967	0.05153	0.04577	0.08784
5	$\Theta_1$	0.03553	0.04567	0.04790	0.04933	0.05153	0.04590	0.08941
6	$\Theta_1$	0.03500	0.04573	0.04790	0.04927	0.05160	0.04590	0.08762
7	$\Theta_1$	0.03353	0.04433	0.04783	0.04967	0.05173	0.04557	0.08828
8	$\Theta_1$	0.03287	0.04447	0.04777	0.04980	0.05173	0.04563	0.08845
9	$\Theta_1$	0.03580	0.04547	0.04790	0.04980	0.05140	0.04650	0.08845
10	$\Theta_1$	0.03340	0.04473	0.04810	0.05007	0.05160	0.04577	0.08765
11	$\Theta_1$	0.03580	0.04593	0.04817	0.05013	0.05167	0.04683	0.08885
12	$\Theta_1$	0.03553	0.04627	0.04797	0.04933	0.05153	0.04643	0.08900
13	$\Theta_1$	0.03467	0.04493	0.04783	0.04980	0.05160	0.04603	0.08890
14	$\Theta_1$	0.03460	0.04507	0.04823	0.05020	0.05167	0.04610	0.08804
15	$\Theta_1$	0.03453	0.04473	0.04783	0.04967	0.05173	0.04603	0.08796
16	$\Theta_1$	0.03327	0.04460	0.04790	0.04993	0.05167	0.04570	0.08720
17	$\Theta_1$	0.03293	0.04460	0.04803	0.04993	0.05153	0.04570	0.08855
18	$\Theta_1$	0.03227	0.04447	0.04777	0.04987	0.05160	0.04557	0.08629
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19	$\Theta_1$	0.03467	0.04420	0.04817	0.04980	0.05147	0.04543	0.08956
20	$\Theta_1$	0.03573	0.04467	0.04790	0.04987	0.05147	0.04577	0.08891
21	$\Theta_1$	0.03407	0.04460	0.04810	0.05007	0.05173	0.04570	0.08797
22	$\Theta_1$	0.03387	0.04540	0.04817	0.05013	0.05173	0.04643	0.08903
23	$\Theta_1$	0.03460	0.04467	0.04797	0.04980	0.05167	0.04583	0.08731
24	$\Theta_1$	0.03353	0.04520	0.04843	0.05013	0.05187	0.04630	0.08952
25	$\Theta_1$	0.03527	0.04487	0.04810	0.04987	0.05173	0.04603	0.08819
26	$\Theta_1$	0.03420	0.04527	0.04823	0.05013	0.05180	0.04630	0.08926
27	$\Theta_1$	0.03487	0.04527	0.04797	0.04993	0.05160	0.04637	0.08924
28	$\Theta_1$	0.03413	0.04513	0.04803	0.05007	0.05187	0.04617	0.08936
29	$\Theta_1$	0.03513	0.04473	0.04790	0.04967	0.05173	0.04597	0.08721
30	$\Theta_1$	0.03453	0.04467	0.04790	0.04993	0.05167	0.04570	0.08782
31	$\Theta_1$	0.03747	0.04500	0.04777	0.04947	0.05167	0.04623	0.08954
32	$\Theta_1$	0.03433	0.04473	0.04797	0.04973	0.05160	0.04590	0.08818
33	$\Theta_1$	0.03627	0.04540	0.04803	0.05000	0.05153	0.04637	0.08860
34	$\Theta_1$	0.03487	0.04520	0.04810	0.05000	0.05187	0.04630	0.08774
35	$\Theta_1$	0.03460	0.04487	0.04810	0.05007	0.05180	0.04597	0.08797
36	$\Theta_1$	0.03500	0.04433	0.04790	0.04973	0.05173	0.04557	0.08743
37	$\Theta_1$	0.03620	0.04553	0.04817	0.04993	0.05173	0.04663	0.08933
38	$\Theta_1$	0.03587	0.04520	0.04790	0.04973	0.05153	0.04630	0.08871
39	$\Theta_1$	0.03327	0.04513	0.04810	0.05013	0.05193	0.04617	0.08778
40	$\Theta_1$	0.03607	0.04613	0.04803	0.04927	0.05187	0.04637	0.08902
41	$\Theta_1$	0.03387	0.04527	0.04810	0.04967	0.05160	0.04543	0.08608

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.03527	0.04473	0.04777	0.04967	0.05153	0.04590	0.08821
43	$\Theta_1$	0.03513	0.04500	0.04797	0.04973	0.05173	0.04617	0.08845
44	$\Theta_1$	0.03593	0.04527	0.04810	0.05027	0.05160	0.04623	0.08842
45	$\Theta_1$	0.03480	0.04480	0.04790	0.04980	0.05173	0.04597	0.08817
46	$\Theta_1$	0.03540	0.04473	0.04817	0.04993	0.05180	0.04590	0.08827
47	$\Theta_1$	0.03667	0.04493	0.04797	0.04967	0.05167	0.04623	0.08869
48	$\Theta_1$	0.03467	0.04553	0.04770	0.04927	0.05147	0.04570	0.08707
49	$\Theta_1$	0.03607	0.04493	0.04810	0.04993	0.05180	0.04603	0.08931
50	$\Theta_1$	0.03593	0.04507	0.04763	0.04947	0.05147	0.04623	0.08838
51	$\Theta_1$	0.03720	0.04620	0.04803	0.04947	0.05140	0.04650	0.08953
52	$\Theta_1$	0.03387	0.04433	0.04790	0.04987	0.05153	0.04543	0.08749
53	$\Theta_1$	0.03527	0.04460	0.04797	0.04973	0.05160	0.04583	0.08847
54	$\Theta_1$	0.03540	0.04473	0.04790	0.04960	0.05173	0.04603	0.08887
55	$\Theta_1$	0.03620	0.04527	0.04803	0.04960	0.05160	0.04650	0.08926
56	$\Theta_1$	0.03387	0.04473	0.04817	0.05020	0.05187	0.04577	0.08759
57	$\Theta_1$	0.03660	0.04627	0.04810	0.04927	0.05173	0.04663	0.08876
58	$\Theta_1$	0.03500	0.04500	0.04797	0.04987	0.05167	0.04610	0.08794
59	$\Theta_1$	0.03673	0.04507	0.04797	0.05000	0.05153	0.04610	0.08885
60	$\Theta_1$	0.03553	0.04527	0.04823	0.05020	0.05160	0.04630	0.08907
61	$\Theta_1$	0.03493	0.04520	0.04817	0.05007	0.05180	0.04623	0.08836

62	$\Theta_1$	0.03447	0.04587	0.04823	0.05000	0.05180	0.04603	0.08935
63	$\Theta_1$	0.03507	0.04447	0.04810	0.04993	0.05160	0.04563	0.08831
64	$\Theta_1$	0.03687	0.04513	0.04810	0.04993	0.05167	0.04630	0.08877
65	$\Theta_1$	0.03380	0.04480	0.04803	0.05013	0.05160	0.04583	0.08782
66	$\Theta_1$	0.03567	0.04520	0.04817	0.05000	0.05167	0.04630	0.08868
67	$\Theta_1$	0.03407	0.04493	0.04817	0.05013	0.05153	0.04590	0.08762
68	$\Theta_1$	0.03673	0.04573	0.04810	0.05007	0.05187	0.04683	0.08922
69	$\Theta_1$	0.03413	0.04440	0.04790	0.04973	0.05167	0.04563	0.08823
70	$\Theta_1$	0.03367	0.04533	0.04817	0.05000	0.05173	0.04643	0.08841
71	$\Theta_1$	0.03467	0.04507	0.04797	0.04980	0.05187	0.04617	0.08862
72	$\Theta_1$	0.03553	0.04607	0.04810	0.04940	0.05167	0.04623	0.08933
73	$\Theta_1$	0.03687	0.04460	0.04777	0.04933	0.05180	0.04603	0.08964
74	$\Theta_1$	0.03500	0.04440	0.04783	0.04953	0.05160	0.04570	0.08777
75	$\Theta_1$	0.03387	0.04527	0.04797	0.04987	0.05173	0.04637	0.08832
76	$\Theta_1$	0.03653	0.04553	0.04817	0.05007	0.05160	0.04663	0.08913
77	$\Theta_1$	0.03587	0.04487	0.04790	0.04967	0.05173	0.04603	0.08886
78	$\Theta_1$	0.03647	0.04527	0.04817	0.04987	0.05173	0.04643	0.08891
79	$\Theta_1$	0.03420	0.04480	0.04790	0.04980	0.05147	0.04563	0.08867
80	$\Theta_1$	0.03527	0.04513	0.04810	0.05007	0.05167	0.04623	0.08919
81	$\Theta_1$	0.03620	0.04547	0.04790	0.05007	0.05173	0.04650	0.08968
82	$\Theta_1$	0.03300	0.04453	0.04810	0.05007	0.05187	0.04563	0.08755
83	$\Theta_1$	0.03453	0.04513	0.04803	0.05013	0.05160	0.04603	0.08829
84	$\Theta_1$	0.03587	0.04473	0.04823	0.05000	0.05167	0.04590	0.08937

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.03573	0.04480	0.04810	0.04987	0.05173	0.04597	0.08826
86	$\Theta_1$	0.03593	0.04580	0.04783	0.04933	0.05173	0.04603	0.08865
87	$\Theta_1$	0.03440	0.04447	0.04783	0.04967	0.05167	0.04570	0.08673
88	$\Theta_1$	0.03400	0.04567	0.04783	0.04927	0.05173	0.04583	0.08758
89	$\Theta_1$	0.03560	0.04480	0.04790	0.04973	0.05147	0.04590	0.08889
90	$\Theta_1$	0.03587	0.04473	0.04797	0.04980	0.05160	0.04590	0.08743
91	$\Theta_1$	0.03593	0.04507	0.04830	0.05013	0.05173	0.04617	0.08877
92	$\Theta_1$	0.03527	0.04493	0.04797	0.04993	0.05173	0.04603	0.08797
93	$\Theta_1$	0.03420	0.04520	0.04830	0.05040	0.05180	0.04597	0.08833
94	$\Theta_1$	0.03947	0.04520	0.04790	0.04967	0.05160	0.04637	0.08893
95	$\Theta_1$	0.03427	0.04460	0.04770	0.04960	0.05153	0.04583	0.08795
96	$\Theta_1$	0.03527	0.04473	0.04797	0.04980	0.05180	0.04590	0.08788
97	$\Theta_1$	0.03433	0.04507	0.04830	0.05027	0.05193	0.04610	0.08866
98	$\Theta_1$	0.03540	0.04493	0.04770	0.04967	0.05167	0.04603	0.08859
99	$\Theta_1$	0.03373	0.04593	0.04810	0.04973	0.05187	0.04623	0.08931
100	$\Theta_1$	0.03647	0.04507	0.04817	0.04993	0.05167	0.04623	0.08872
All	$\Theta_1$	0.01487	0.01893	0.02117	0.02280	0.02620	0.02083	0.09985

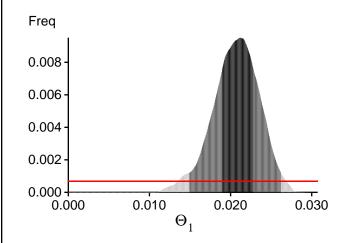
Citation suggestions:

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	-15920.81	-15398.06	-15394.58	-15468.89
2	-17535.59	-16629.21	-16598.43	-16652.17
3	-16495.21	-15724.72	-15695.46	-15760.57
4	-16115.04	-15472.52	-15460.51	-15524.49
5	-16908.90	-15900.26	-15830.37	-15891.40
6	-15928.07	-15372.03	-15386.05	-15443.42
7	-18387.76	-16740.90	-16556.13	-16617.88
8	-16566.08	-15767.43	-15728.46	-15797.47
9	-16370.47	-15558.15	-15523.96	-15581.19
10	-15864.89	-15142.11	-15113.15	-15178.46
11	-15976.65	-15455.46	-15472.13	-15535.39
12	-18529.28	-16524.64	-16265.36	-16330.68
13	-16560.47	-15746.81	-15711.58	-15766.57
14	-16385.70	-15748.20	-15752.10	-15812.54
15	-16183.59	-15548.59	-15549.10	-15604.98
16	-16137.22	-15489.11	-15492.01	-15546.08
17	-16250.29	-15525.15	-15512.37	-15568.16
18	-15326.33	-15003.71	-15029.09	-15102.54
19	-16612.27	-15950.53	-15952.48	-16007.22
20	-17676.64	-16262.70	-16127.93	-16180.22
21	-17497.94	-16378.34	-16292.84	-16353.86
22	-16753.91	-15811.29	-15744.14	-15814.62
23	-15331.87	-14901.13	-14932.11	-14991.28
24	-18693.16	-16901.43	-16699.70	-16749.53
25	-17165.55	-16019.81	-15916.77	-15984.29
26	-16565.74	-15912.70	-15919.54	-15973.11
27	-15997.90	-15447.19	-15465.87	-15516.05
28	-16811.94	-15889.85	-15840.57	-15903.01
29	-15150.78	-14754.79	-14780.36	-14846.56

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 19:32:45]

30	-17328.61	-15991.00	-15857.12	-15915.33
31	-18027.66	-16292.37	-16089.32	-16149.16
32	-15982.00	-15286.65	-15260.55	-15331.16
33	-16612.80	-15849.77	-15829.52	-15882.57
34	-15081.38	-14727.28	-14765.20	-14826.56
35	-15443.03	-14905.75	-14911.80	-14974.80
36	-16132.27	-15530.42	-15540.08	-15592.18
37	-16664.58	-16003.53	-16008.48	-16058.50
38	-15826.94	-15161.49	-15153.48	-15208.87
39	-15433.53	-14791.43	-14774.61	-14840.27
40	-16717.52	-15749.37	-15684.78	-15742.38
41	-14506.72	-14269.49	-14312.26	-14388.95
42	-18993.14	-16872.41	-14763.20	-16662.50
43	-16919.05	-16062.40	-14789.32	-16079.01
44	-16295.08	-15466.48	-15273.63	-15487.46
45	-16909.27	-16285.70	-15135.84	-16345.93
46	-17008.85	-15902.20	-15822.51	-15876.06
47	-16054.95	-15282.87	-14920.62	-15313.84
48	-16428.30	-15840.63	-15840.08	-15909.97
49	-16114.29	-15444.74	-15443.28	-15494.30
50	-15898.42	-15310.20	-15318.20	-15372.75
51	-17620.02	-16586.23	-16282.67	-16578.76
52	-16203.11	-15594.87	-14935.40	-15657.19
53	-17361.94	-16100.83	-15866.44	-16047.18
54	-16332.89	-15541.89	-15514.56	-15572.85
55	-16140.13	-15610.88	-15521.62	-15689.79
56	-17658.89	-16479.65	-15699.77	-16447.78
57	-17489.60	-16092.74	-15446.61	-16009.24
58	-14612.96	-14365.66	-14420.68	-14489.29
59	-15552.70	-15128.65	-15163.26	-15218.43
60	-18558.58	-16604.95	-16122.85	-16416.01
61	-17066.49	-16005.49	-15686.16	-15985.86
62	-17141.69	-16271.27	-15552.45	-16287.02
63	-18242.80	-16815.28	-15156.57	-16736.12
64	-16705.20	-15531.65	-15430.34	-15492.90
65	-16482.60	-15677.82	-15540.67	-15705.65
66	-18294.19	-16501.43	-15455.95	-16347.44
67	-15656.59	-15148.79	-15166.66	-15224.49
68	-20189.33	-17777.58	-15961.16	-17518.27
69	-16882.17	-15920.41	-15853.86	-15915.37
70	-17154.06	-16108.49	-15924.47	-16092.09
71	-15884.16	-15393.18	-15426.84	-15481.39
72	-16755.11	-16037.35	-15741.48	-16087.95
73	-17522.28	-16508.94	-15416.11	-16500.04
74	-16970.44	-15741.97	-15494.33	-15687.44

All	-1675448.38	-1581414.68	-1552191.46	-1581616.52
100	-10022.29	-10004.21	-10101.44	-10400.30
99 100	-18679.65 -16022.29	-17131.65 -15384.21	-16033.81 -15181.44	-17039.95 -15438.35
98	-15848.28	-15311.97	-15324.65	-15385.61
97	-15801.93	-15320.39	-15352.42	-15406.03
96	-16625.49	-15909.95	-15908.05	-15959.56
95 06	-16260.56	-15531.51	-15514.52	-15570.34
94	-17250.62	-16061.80	-15968.41	-16019.36
93	-16433.56	-15630.57	-15595.59	-15654.80
92	-17596.21	-16034.49	-15518.42	-15918.24
91	-16188.35	-15549.85	-15552.89	-15604.58
90	-15711.27	-15053.43	-15033.74	-15100.07
89	-16466.80	-15887.74	-15630.93	-15959.12
88	-16843.17	-15693.67	-15428.40	-15658.54
87	-18948.23	-16346.90	-15427.17	-16037.75
86	-16684.81	-15834.67	-15142.28	-15855.49
85 86	-16061.05	-15376.24	-15369.50	-15421.92
84	-19658.78	-17559.60	-15269.74	-17358.27
83	-15273.37	-14857.55	-14426.05	-14950.10
82	-15139.87	-14696.48	-14719.21	-14781.07
81	-18540.00	-17131.11	-14333.86	-17061.08
80	-17171.26	-16244.24	-16209.64	-16258.32
79	-17238.10	-15955.51	-15838.33	-15896.46
78 70	-16917.01	-16201.06	-16191.47	-16248.44
77	-16539.55	-15618.04	-15560.25	-15619.49
76 77	-19031.47	-17171.17	-15832.77	-17012.64
75 70	-18145.34	-16556.86	-15994.97	-16442.24

- (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 = 185.008553]

#### 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	364461552/399970288 67998078/1600029712	0.91122 0.04250

# MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$	0.37861	11712621.02
Genealogies	0.65962	5528726.36

# 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