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 Sat Aug 12 22:32:36 2017

Program finished at Sun Aug 13 00:34:15 2017 [Runtime:0000:02:01:39]



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) 3159841511

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 Θ_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.0.4 NO

Haplotyping is turned on:

Output file: outfile_0.4_0.9

Posterior distribution raw histogram file: bayesfile Raw data from the MCMC run: bayesallfile_0.4_0.9

Print data: No

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

Data summary

Data file: infile.0.4
Datatype: Sequence data
Number of loci: 100

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Mutation	model:			
Locus Su	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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[Basefreq: =0.25]

Jukes-Cantor

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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	
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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	
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32	1	1	1.000	1.000	1.000	
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67	1	1	1.000	1.000	1.000	
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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	
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97	1	1	1.000	1.000	1.000	
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99	1	1	1.000	1.000	1.000	
100	1	1	1.000	1.000	1.000	
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Total of all populations	1	10	
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100	10

Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
2	Θ_1	0.00000	0.00053	0.00150	0.00240	0.00480	0.00197	0.00171
3	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
4	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
5	Θ_1	0.00000	0.00100	0.00217	0.00313	0.00533	0.00243	0.00234
6	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
7	Θ_1	0.00000	0.00000	0.00070	0.00133	0.00293	0.00130	0.00075
8	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
9	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
10	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
11	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
12	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
13	Θ_1	0.00000	0.00027	0.00110	0.00187	0.00347	0.00157	0.00116
14	Θ_1	0.00000	0.00013	0.00090	0.00160	0.00313	0.00143	0.00094
15	Θ_1	0.00000	0.00000	0.00063	0.00127	0.00287	0.00130	0.00068
16	Θ_1	0.00000	0.00000	0.00063	0.00127	0.00280	0.00130	0.00067
17	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
18	Θ_1	0.00000	0.00093	0.00223	0.00347	0.00773	0.00283	0.00300

19	Θ_1	0.00000	0.00000	0.00057	0.00120	0.00267	0.00123	0.00056
20	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
21	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
22	Θ_1	0.00000	0.00093	0.00217	0.00320	0.00633	0.00257	0.00257
23	Θ_1	0.00000	0.00000	0.00063	0.00127	0.00287	0.00130	0.00068
24	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
25	Θ_1	0.00000	0.00000	0.00077	0.00140	0.00293	0.00137	0.00078
26	Θ_1	0.00000	0.00007	0.00077	0.00140	0.00293	0.00130	0.00076
27	Θ_1	0.00000	0.00027	0.00103	0.00180	0.00333	0.00150	0.00106
28	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
29	Θ_1	0.00000	0.00007	0.00077	0.00147	0.00300	0.00137	0.00081
30	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
31	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
32	Θ_1	0.00000	0.00000	0.00063	0.00120	0.00280	0.00123	0.00064
33	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
34	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
35	Θ_1	0.00000	0.00073	0.00177	0.00267	0.00460	0.00203	0.00185
36	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
37	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
38	Θ_1	0.00000	0.00000	0.00063	0.00120	0.00280	0.00123	0.00065
39	Θ_1	0.00000	0.00053	0.00150	0.00240	0.00473	0.00197	0.00170
40	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
41	Θ_1	0.00000	0.00013	0.00083	0.00153	0.00307	0.00143	0.00088

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
43	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
44	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
45	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
46	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
47	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
48	Θ_1	0.00000	0.00000	0.00063	0.00127	0.00280	0.00130	0.00066
49	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
50	Θ_1	0.00000	0.00147	0.00303	0.00467	0.01007	0.00383	0.00424
51	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
52	Θ_1	0.00007	0.00047	0.00103	0.00160	0.00193	0.00157	0.00114
53	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
54	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
55	Θ_1	0.00000	0.00007	0.00077	0.00140	0.00293	0.00137	0.00076
56	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
57	Θ_1	0.00000	0.00007	0.00083	0.00147	0.00300	0.00137	0.00082
58	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
59	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
60	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
61	Θ_1	0.00000	0.00100	0.00237	0.00353	0.00753	0.00290	0.00303

62	Θ_1	0.00000	0.00000	0.00070	0.00133	0.00287	0.00130	0.00071
63	Θ_1	0.00000	0.00000	0.00063	0.00120	0.00280	0.00123	0.00064
64	Θ_1	0.00000	0.00020	0.00097	0.00167	0.00320	0.00143	0.00097
65	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
66	Θ_1	0.00000	0.00000	0.00063	0.00127	0.00280	0.00130	0.00065
67	Θ_1	0.00000	0.00013	0.00083	0.00153	0.00307	0.00137	0.00086
68	Θ_1	0.00000	0.00013	0.00090	0.00153	0.00307	0.00143	0.00089
69	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
70	Θ_1	0.00000	0.00113	0.00237	0.00347	0.00620	0.00277	0.00272
71	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
72	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
73	Θ_1	0.00000	0.00020	0.00097	0.00167	0.00320	0.00143	0.00097
74	Θ_1	0.00000	0.00000	0.00057	0.00120	0.00267	0.00123	0.00056
75	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
76	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
77	Θ_1	0.00000	0.00040	0.00123	0.00207	0.00367	0.00163	0.00129
78	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
79	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
80	Θ_1	0.00000	0.00047	0.00143	0.00227	0.00447	0.00183	0.00157
81	Θ_1	0.00000	0.00000	0.00070	0.00133	0.00293	0.00130	0.00075
82	Θ_1	0.00000	0.00000	0.00070	0.00133	0.00287	0.00130	0.00072
83	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
84	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	Θ_1	0.00000	0.00000	0.00063	0.00120	0.00280	0.00123	0.00064
86	Θ_1	0.00000	0.00000	0.00063	0.00127	0.00280	0.00130	0.00066
87	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
88	Θ_1	0.00007	0.00213	0.00357	0.00500	0.00880	0.00403	0.00427
89	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
90	Θ_1	0.00000	0.00020	0.00097	0.00167	0.00327	0.00143	0.00099
91	Θ_1	0.00000	0.00000	0.00057	0.00093	0.00093	0.00123	0.00056
92	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
93	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
94	Θ_1	0.00000	0.00000	0.00063	0.00127	0.00280	0.00130	0.00065
95	Θ_1	0.00000	0.00007	0.00077	0.00147	0.00300	0.00137	0.00079
96	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
97	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
98	Θ_1	0.00000	0.00000	0.00050	0.00120	0.00267	0.00123	0.00056
99	Θ_1	0.00000	0.00000	0.00077	0.00140	0.00300	0.00137	0.00079
100	Θ_1	0.00000	0.00000	0.00057	0.00120	0.00267	0.00123	0.00056
All	Θ_1	0.00000	0.00000	0.00043	0.00113	0.00247	0.00117	0.00045

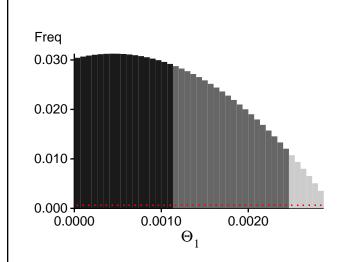
Citation suggestions:

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.						
2 2 3 3 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4 4						

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	-14021.46	-13755.83	-13773.83	-13868.23
2	-14538.79	-14177.23	-14197.83	-14277.03
3	-14020.36	-13756.62	-13775.70	-13868.97
4	-14021.43	-13756.95	-13776.07	-13868.93
5	-14163.95	-13896.77	-13935.66	-14011.19
6	-14019.76	-13756.16	-13774.70	-13868.40
7	-14049.47	-13783.07	-13803.36	-13894.08
8	-14014.37	-13752.19	-13771.67	-13865.40
9	-14017.96	-13754.66	-13774.13	-13867.07
10	-14019.86	-13755.66	-13774.60	-13867.86
11	-14019.67	-13755.50	-13774.82	-13868.88
12	-14021.19	-13756.97	-13776.17	-13869.35
13	-14166.33	-13901.67	-13934.86	-14018.78
14	-14130.94	-13842.49	-13868.50	-13952.10
15	-14034.66	-13769.55	-13791.34	-13882.25
16	-14031.06	-13767.71	-13789.11	-13880.31
17	-14022.23	-13756.92	-13775.40	-13868.74
18	-14821.46	-14510.53	-14549.09	-14622.38
19	-14019.93	-13755.29	-13774.32	-13867.72
20	-14018.44	-13756.11	-13775.18	-13869.81
21	-14020.97	-13756.66	-13776.06	-13869.26
22	-16295.90	-15712.92	-15717.24	-15786.00
23	-14030.56	-13766.92	-13788.63	-13879.81
24	-14021.79	-13756.44	-13775.56	-13869.67
25	-14078.04	-13804.08	-13829.08	-13915.58
26	-14049.80	-13782.87	-13802.52	-13895.21
27	-14179.45	-13888.71	-13916.13	-13998.59
28	-14018.91	-13756.22	-13775.36	-13868.89
29	-14043.16	-13779.33	-13803.18	-13892.31

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 22:32:36]

30	-14020.23	-13756.42	-13775.53	-13868.99
31	-14020.62	-13756.59	-13776.00	-13869.30
32	-14034.11	-13770.08	-13789.53	-13881.25
33	-14020.02	-13756.32	-13775.44	-13868.69
34	-14019.16	-13756.18	-13774.71	-13869.69
35	-14115.94	-13855.65	-13891.24	-13971.73
36	-14019.39	-13756.70	-13775.57	-13869.23
37	-14019.47	-13756.44	-13775.68	-13869.37
38	-14031.96	-13767.81	-13787.74	-13880.75
39	-16962.82	-16061.29	-16004.77	-16077.36
40	-14017.65	-13753.10	-13771.89	-13865.61
41	-14114.55	-13831.12	-13856.18	-13941.78
42	-14021.94	-13756.55	-13775.43	-13868.31
43	-14018.52	-13755.99	-13775.30	-13868.80
44	-14018.29	-13754.56	-13773.77	-13868.99
45	-14017.02	-13755.32	-13774.56	-13871.47
46	-14020.69	-13756.49	-13775.71	-13868.90
47	-14017.98	-13754.01	-13773.05	-13866.05
48	-14034.84	-13769.67	-13789.07	-13881.52
49	-14018.52	-13753.60	-13772.29	-13866.22
50	-24906.02	-23598.06	-23565.56	-23625.99
51	-14020.65	-13757.00	-13776.35	-13869.52
52	-24967.23	-21250.90	-20713.59	-20786.45
53	-14017.38	-13753.72	-13772.88	-13866.44
54	-14019.83	-13755.24	-13773.50	-13867.22
55	-14061.59	-13797.43	-13821.33	-13909.44
56	-14020.85	-13756.37	-13775.58	-13868.74
57	-14082.62	-13814.36	-13840.30	-13927.93
58	-14021.51	-13756.60	-13775.27	-13868.97
59	-14017.56	-13754.75	-13774.40	-13867.51
60	-14020.25	-13756.30	-13775.51	-13868.52
61	-15618.48	-15030.11	-15027.76	-15096.30
62	-14058.37	-13787.12	-13810.03	-13898.43
63	-14038.18	-13769.86	-13788.47	-13880.06
64	-14147.29	-13854.48	-13880.16	-13962.60
65	-14018.36	-13754.38	-13773.73	-13866.54
66	-14032.35	-13768.29	-13788.45	-13879.61
67	-14064.05	-13797.93	-13822.82	-13911.31
68	-14112.56	-13830.10	-13856.16	-13940.13
69	-14022.70	-13757.00	-13775.59	-13869.50
70	-14268.19	-13994.92	-14032.72	-14109.84
71	-14018.66	-13755.54	-13775.04	-13869.56
72	-14020.44	-13756.26	-13775.54	-13868.63
73	-14104.43	-13839.93	-13870.03	-13952.49
74	-14018.75	-13755.67	-13775.06	-13868.84

75	-14020.01	-13755.70	-13775.15	-13868.44
76	-14018.63	-13755.34	-13774.80	-13868.05
77	-14223.26	-13935.61	-13964.71	-14046.09
78	-14017.64	-13754.24	-13773.72	-13866.81
79	-14020.63	-13755.43	-13774.11	-13867.40
80	-18259.51	-16897.44	-16762.04	-16835.48
81	-14045.48	-13779.96	-13801.57	-13893.36
82	-14052.65	-13787.22	-13810.44	-13900.64
83	-14019.53	-13754.98	-13773.97	-13867.01
84	-14020.06	-13755.92	-13774.83	-13869.00
85	-14033.35	-13767.65	-13787.03	-13882.15
86	-14031.43	-13767.35	-13787.08	-13878.76
87	-14017.02	-13753.23	-13772.55	-13865.76
88	-14346.43	-14081.27	-14132.64	-14201.26
89	-14020.93	-13757.04	-13775.93	-13869.53
90	-14065.53	-13800.25	-13826.52	-13911.04
91	-14022.82	-13757.26	-13776.15	-13869.82
92	-14022.18	-13756.90	-13775.96	-13869.46
93	-14018.91	-13754.64	-13773.75	-13867.26
94	-14028.23	-13764.98	-13784.72	-13876.36
95	-14045.30	-13781.28	-13804.08	-13892.91
96	-14018.42	-13755.50	-13774.02	-13868.50
97	-14017.92	-13754.67	-13773.99	-13867.45
98	-14018.57	-13755.57	-13774.65	-13869.45
99	-14058.10	-13790.95	-13815.74	-13902.88
100	-14018.21	-13755.45	-13774.51	-13867.73
All	-1438216.11	-1404605.49	-1405885.46	-1414831.39

- (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 = 438.546760]

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
Θ_1 Genealogies	73624666/399996510 1017855515/1600003490	0.18406 0.63616

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
Θ_1	0.03633	9306041.99
Genealogies	0.04447	9170247.31

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