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 21:48:57 2017

Program finished at Sun Aug 13 23:19:42 2017 [Runtime:0000:01:30:45]



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) 3373159966

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.1.0 NO

Haplotyping is turned on:

Output file: outfile_1.0_0.7 Posterior distribution raw histogram file: bayesfile

Raw data from the MCMC run: bayesallfile_1.0_0.7

Print data: No

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

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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99	1	Jukes-Cantor	[Basefreq: =0.25]	
100	1	Jukes-Cantor	[Basefreq: =0.25]	
Sites per	locus			
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Locus	Sites
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Site rate	e variation and probat	oilities:			
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 8 9	1 1	1	1.000	1.000	1.000	
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	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	
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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	
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69	1	1	1.000	1.000	1.000	
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73	1	1	1.000	1.000	1.000	
74	1	1	1.000	1.000	1.000	
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76	1	1	1.000	1.000	1.000	
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82	1	1	1.000	1.000	1.000	
83	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	
88	1	1	1.000	1.000	1.000	
89	1	1	1.000	1.000	1.000	
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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	
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Total of all populations	1	10	
	2	10	
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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	Θ_1	0.02793	0.04180	0.04763	0.04940	0.05140	0.04277	0.06943
2	Θ_1	0.02720	0.04207	0.04763	0.04907	0.05133	0.04223	0.06624
3	Θ_1	0.03180	0.04353	0.04790	0.04980	0.05153	0.04470	0.07797
4	Θ_1	0.03080	0.04393	0.04777	0.04933	0.05147	0.04417	0.07422
5	Θ_1	0.02927	0.04233	0.04777	0.04953	0.05153	0.04370	0.07503
6	Θ_1	0.02467	0.03913	0.04750	0.04900	0.05107	0.04050	0.06210
7	Θ_1	0.03073	0.04287	0.04777	0.04960	0.05147	0.04417	0.07574
8	Θ_1	0.02967	0.04133	0.04763	0.04967	0.05147	0.04357	0.07217
9	Θ_1	0.02047	0.03527	0.04237	0.04693	0.05060	0.03703	0.05127
10	Θ_1	0.02527	0.04053	0.04763	0.04907	0.05113	0.04103	0.06265
11	Θ_1	0.02933	0.04247	0.04777	0.04967	0.05153	0.04377	0.07568
12	Θ_1	0.03047	0.04300	0.04763	0.04953	0.05147	0.04423	0.07831
13	Θ_1	0.02553	0.04127	0.04763	0.04907	0.05120	0.04143	0.06527
14	Θ_1	0.02753	0.03587	0.04757	0.05020	0.05120	0.04243	0.06788
15	Θ_1	0.03240	0.04473	0.04790	0.04940	0.05160	0.04497	0.08025
16	Θ_1	0.02953	0.04247	0.04770	0.04953	0.05140	0.04377	0.07413
17	Θ_1	0.02707	0.03847	0.04757	0.04967	0.05127	0.04203	0.06535
18	Θ_1	0.02867	0.04220	0.04770	0.04960	0.05147	0.04350	0.07426

19	Θ_1	0.02393	0.03987	0.04750	0.04860	0.05100	0.04003	0.05919
20	Θ_1	0.03067	0.04400	0.04783	0.04953	0.05160	0.04423	0.07792
21	Θ_1	0.02813	0.04153	0.04763	0.04947	0.05140	0.04283	0.07084
22	Θ_1	0.01967	0.03460	0.04030	0.04607	0.05053	0.03657	0.05033
23	Θ_1	0.03040	0.04367	0.04777	0.04933	0.05153	0.04397	0.07449
24	Θ_1	0.02607	0.04073	0.04757	0.04933	0.05127	0.04217	0.06767
25	Θ_1	0.03073	0.04293	0.04777	0.04967	0.05153	0.04417	0.07475
26	Θ_1	0.02720	0.04080	0.04757	0.04927	0.05127	0.04223	0.06606
27	Θ_1	0.02207	0.03760	0.04750	0.04880	0.05100	0.03923	0.05898
28	Θ_1	0.03267	0.04367	0.04783	0.04973	0.05153	0.04483	0.07845
29	Θ_1	0.02947	0.04333	0.04763	0.04933	0.05140	0.04350	0.07305
30	Θ_1	0.02100	0.03760	0.04583	0.04807	0.05073	0.03783	0.05288
31	Θ_1	0.02900	0.04187	0.04770	0.04947	0.05140	0.04323	0.07073
32	Θ_1	0.02993	0.03993	0.04777	0.05027	0.05153	0.04390	0.07518
33	Θ_1	0.03093	0.04300	0.04783	0.04973	0.05147	0.04423	0.07570
34	Θ_1	0.03353	0.04420	0.04777	0.04973	0.05153	0.04537	0.08160
35	Θ_1	0.02433	0.04047	0.04750	0.04893	0.05120	0.04070	0.06321
36	Θ_1	0.02927	0.04220	0.04770	0.04960	0.05147	0.04343	0.07334
37	Θ_1	0.02993	0.04260	0.04770	0.04960	0.05153	0.04390	0.07598
38	Θ_1	0.03140	0.04360	0.04783	0.04980	0.05160	0.04477	0.07969
39	Θ_1	0.02960	0.04253	0.04777	0.04967	0.05147	0.04377	0.07524
40	Θ_1	0.02740	0.04067	0.04757	0.04927	0.05127	0.04210	0.06693
41	Θ_1	0.02420	0.03940	0.04757	0.04920	0.05120	0.04090	0.06490

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	Θ_1	0.03273	0.04380	0.04797	0.04980	0.05160	0.04497	0.07945
43	Θ_1	0.02473	0.03927	0.04757	0.04913	0.05113	0.04070	0.06216
44	Θ_1	0.03160	0.04333	0.04783	0.04967	0.05147	0.04457	0.07716
45	Θ_1	0.02880	0.04273	0.04770	0.04927	0.05140	0.04310	0.07139
46	Θ_1	0.02433	0.03940	0.04757	0.04913	0.05120	0.04083	0.06390
47	Θ_1	0.02793	0.04240	0.04763	0.04927	0.05140	0.04270	0.06906
48	Θ_1	0.03167	0.04407	0.04777	0.04933	0.05153	0.04437	0.07720
49	Θ_1	0.01833	0.02987	0.03630	0.04547	0.05033	0.03523	0.04721
50	Θ_1	0.02847	0.04187	0.04770	0.04960	0.05140	0.04317	0.07191
51	Θ_1	0.02747	0.04127	0.04770	0.04953	0.05133	0.04250	0.06876
52	Θ_1	0.03287	0.04407	0.04783	0.04980	0.05153	0.04523	0.07942
53	Θ_1	0.02887	0.04227	0.04770	0.04960	0.05147	0.04350	0.07240
54	Θ_1	0.02440	0.04033	0.04757	0.04913	0.05120	0.04077	0.06379
55	Θ_1	0.02947	0.04147	0.04763	0.04967	0.05133	0.04343	0.07160
56	Θ_1	0.03040	0.04293	0.04777	0.04967	0.05153	0.04417	0.07626
57	Θ_1	0.02260	0.03820	0.04750	0.04900	0.05107	0.03977	0.05974
58	Θ_1	0.02340	0.03967	0.04757	0.04873	0.05107	0.03983	0.05962
59	Θ_1	0.02167	0.03053	0.04750	0.04967	0.05087	0.03877	0.05814
60	Θ_1	0.03113	0.04340	0.04790	0.04980	0.05160	0.04457	0.07919
61	Θ_1	0.03053	0.03820	0.04777	0.05060	0.05153	0.04423	0.07600

62	Θ_1	0.03067	0.04307	0.04783	0.04973	0.05153	0.04423	0.07685
63	Θ_1	0.01907	0.03033	0.03617	0.04487	0.05027	0.03543	0.04711
64	Θ_1	0.02973	0.04240	0.04777	0.04960	0.05147	0.04363	0.07400
65	Θ_1	0.03387	0.04467	0.04803	0.05000	0.05167	0.04570	0.08353
66	Θ_1	0.03107	0.04313	0.04777	0.04967	0.05153	0.04437	0.07763
67	Θ_1	0.02873	0.04227	0.04770	0.04967	0.05147	0.04350	0.07318
68	Θ_1	0.02953	0.04247	0.04777	0.04967	0.05147	0.04370	0.07295
69	Θ_1	0.02940	0.04267	0.04777	0.04960	0.05147	0.04390	0.07650
70	Θ_1	0.03367	0.04440	0.04790	0.04980	0.05160	0.04550	0.08301
71	Θ_1	0.02960	0.04227	0.04770	0.04953	0.05153	0.04363	0.07199
72	Θ_1	0.03107	0.04307	0.04777	0.04960	0.05153	0.04437	0.07692
73	Θ_1	0.02287	0.03893	0.04750	0.04887	0.05107	0.03963	0.06039
74	Θ_1	0.03060	0.04387	0.04790	0.04967	0.05153	0.04430	0.07844
75	Θ_1	0.03133	0.04307	0.04783	0.04967	0.05153	0.04430	0.07484
76	Θ_1	0.02447	0.04073	0.04757	0.04900	0.05113	0.04090	0.06486
77	Θ_1	0.02933	0.04247	0.04777	0.04967	0.05153	0.04370	0.07513
78	Θ_1	0.02660	0.04060	0.04757	0.04940	0.05127	0.04197	0.06526
79	Θ_1	0.02567	0.04113	0.04757	0.04913	0.05127	0.04163	0.06648
80	Θ_1	0.02653	0.04093	0.04770	0.04947	0.05133	0.04223	0.07058
81	Θ_1	0.03200	0.04353	0.04783	0.04967	0.05153	0.04477	0.07911
82	Θ_1	0.02747	0.04100	0.04763	0.04933	0.05133	0.04237	0.07034
83	Θ_1	0.03107	0.03260	0.04770	0.05127	0.05147	0.04423	0.07655
84	Θ_1	0.03207	0.04340	0.04783	0.04973	0.05153	0.04463	0.07707

$\begin{array}{cccccccccccccccccccccccccccccccccccc$									
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	85	Θ_1	0.03080	0.04393	0.04777	0.04933	0.05147	0.04410	0.07476
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	86	Θ_1	0.02487	0.04067	0.04757	0.04913	0.05127	0.04143	0.06793
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	87	Θ_1	0.02713	0.04133	0.04763	0.04947	0.05140	0.04263	0.07238
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	88	Θ_1	0.03173	0.04340	0.04783	0.04973	0.05160	0.04463	0.07719
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	89	Θ_1	0.02487	0.04033	0.04757	0.04880	0.05113	0.04057	0.06052
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	90	Θ_1	0.02313	0.03800	0.04750	0.04873	0.05093	0.03943	0.05681
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	91	Θ_1	0.03113	0.04373	0.04770	0.04933	0.05140	0.04403	0.07485
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	92	Θ_1	0.02867	0.04200	0.04777	0.04953	0.05147	0.04330	0.07269
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	93	Θ_1	0.02427	0.04093	0.04757	0.04920	0.05127	0.04117	0.06687
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	94	Θ_1	0.03167	0.04320	0.04770	0.04960	0.05153	0.04443	0.07663
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	95	Θ_1	0.03180	0.04360	0.04790	0.04980	0.05153	0.04477	0.07745
$\begin{array}{cccccccccccccccccccccccccccccccccccc$	96	Θ_1	0.03227	0.04373	0.04783	0.04960	0.05160	0.04497	0.08115
99 Θ_1 0.03267 0.04407 0.04797 0.04980 0.05160 0.04523 0.080 100 Θ_1 0.03000 0.04247 0.04770 0.04953 0.05147 0.04377 0.075	97	Θ_1	0.02993	0.04347	0.04763	0.04920	0.05140	0.04363	0.07364
100 Θ_1 0.03000 0.04247 0.04770 0.04953 0.05147 0.04377 0.075	98	Θ_1	0.03300	0.04507	0.04790	0.04947	0.05153	0.04523	0.08121
	99	Θ_1	0.03267	0.04407	0.04797	0.04980	0.05160	0.04523	0.08012
All Θ_1 0.00287 0.00473 0.00643 0.00820 0.00947 0.00730 0.080	100	Θ_1	0.03000	0.04247	0.04770	0.04953	0.05147	0.04377	0.07561
•	All	Θ_1	0.00287	0.00473	0.00643	0.00820	0.00947	0.00730	0.08077

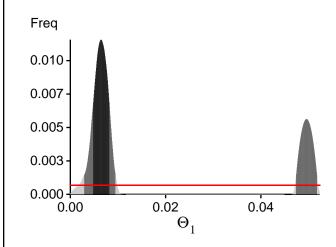
Citation suggestions:

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	-15688.14	-15180.73	-15216.52	-15263.23
2	-16159.80	-15583.21	-15608.41	-15656.93
3	-17262.28	-16480.85	-16484.01	-16525.85
4	-16916.99	-16043.49	-16020.75	-16066.20
5	-15999.37	-15554.32	-15607.11	-15653.81
6	-15269.57	-14891.92	-14946.29	-14995.71
7	-18100.87	-16683.49	-16564.36	-16609.79
8	-17178.66	-16573.73	-16607.61	-16654.25
9	-15064.56	-14726.20	-14780.82	-14837.97
10	-15362.99	-14967.96	-15019.19	-15070.04
11	-17798.89	-16848.86	-16821.90	-16865.97
12	-17140.69	-16574.81	-16619.46	-16665.72
13	-16795.67	-16019.01	-16011.14	-16059.96
14	-16166.14	-15536.98	-15552.31	-15600.28
15	-17257.52	-16464.51	-16465.18	-16507.83
16	-16140.63	-15626.05	-15666.79	-15712.26
17	-16169.69	-15528.82	-15540.60	-15590.65
18	-17958.45	-16751.64	-16672.28	-16717.64
19	-15253.34	-14839.27	-14883.46	-14936.12
20	-16272.07	-15747.48	-15789.88	-15837.68
21	-15550.49	-15166.14	-15223.80	-15272.96
22	-14875.28	-14593.45	-14656.60	-14713.70
23	-15742.64	-15372.47	-15437.93	-15482.82
24	-16442.02	-15599.11	-15573.30	-15621.75
25	-17152.09	-16193.16	-16155.14	-16200.53
26	-15643.55	-15313.36	-15383.18	-15431.87
27	-15333.06	-14957.84	-15012.54	-15063.64
28	-16703.40	-16168.63	-16215.09	-16257.92
29	-16259.40	-15681.58	-15710.85	-15756.44

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 21:48:57]

30	-15285.32	-14862.21	-14903.18	-14960.25
31	-17825.60	-16739.09	-16681.13	-16728.12
32	-16455.45	-15794.22	-15809.65	-15854.74
33	-16133.00	-15627.44	-15671.11	-15715.27
34	-17863.69	-17188.66	-17221.78	-17262.81
35	-15440.59	-15055.72	-15109.83	-15160.13
36	-15932.45	-15432.13	-15473.70	-15519.29
37	-16450.72	-15879.78	-15914.69	-15957.78
38	-16832.37	-16203.75	-16234.95	-16276.41
39	-16056.70	-15529.41	-15567.70	-15612.40
40	-15957.16	-15416.23	-15445.74	-15495.40
41	-15694.34	-15294.30	-15350.41	-15399.18
42	-16844.38	-16158.27	-16174.19	-16218.30
43	-15817.57	-15256.54	-15278.88	-15330.61
44	-16477.66	-15928.23	-15967.84	-16011.99
45	-16656.23	-15892.02	-15887.38	-15934.10
46	-16168.50	-15549.16	-15565.13	-15614.61
47	-17673.97	-16632.33	-16581.35	-16629.71
48	-16082.43	-15575.10	-15619.17	-15662.49
49	-15030.02	-14747.01	-14808.94	-14866.72
50	-16631.81	-15973.44	-15989.42	-16037.16
51	-17323.82	-16609.53	-16621.58	-16670.06
52	-18265.90	-17126.82	-17068.70	-17114.88
53	-17194.80	-16213.77	-16171.83	-16218.90
54	-15415.07	-15037.74	-15092.79	-15143.75
55	-16082.31	-15637.54	-15691.01	-15737.59
56	-16601.22	-15876.65	-15881.31	-15926.18
57	-15096.93	-14760.64	-14819.54	-14871.64
58	-14951.97	-14690.03	-14761.57	-14814.23
59	-15322.37	-14958.86	-15014.31	-15065.72
60	-17820.84	-16823.75	-16789.47	-16833.52
61	-17232.07	-16599.51	-16630.45	-16674.74
62	-16259.53	-15837.17	-15900.31	-15943.91
63	-14943.21	-14629.62	-14687.33	-14743.08
64	-18954.99	-17101.61	-16901.10	-16947.70
65	-17873.66	-16925.13	-16905.06	-16943.14
66	-17190.14	-16286.82	-16263.07	-16306.73
67	-16926.83	-16115.04	-16104.48	-16149.54
68	-17117.16	-16330.51	-16326.91	-16373.32
69	-16920.89	-16204.97	-16216.01	-16258.71
70	-17798.82	-16903.17	-16892.38	-16931.79
71	-17176.17	-16068.81	-15999.36	-16046.67
72	-17333.55	-16453.22	-16435.42	-16479.06
73	-15485.92	-15071.99	-15119.97	-15171.54
74	-16779.08	-16166.37	-16198.44	-16240.18

75	-17702.56	-16550.32	-16480.25	-16525.00
76	-16869.61	-16153.08	-16156.89	-16207.12
77	-15961.42	-15479.96	-15526.53	-15572.53
78	-16411.19	-15738.40	-15746.35	-15795.81
79	-15865.50	-15383.05	-15421.37	-15472.40
80	-15737.13	-15389.32	-15458.93	-15505.37
81	-18084.59	-16924.37	-16859.55	-16903.02
82	-15671.10	-15233.37	-15282.19	-15329.48
83	-19130.53	-17442.03	-17280.00	-17325.32
84	-16821.69	-16279.06	-16324.41	-16368.45
85	-16142.48	-15665.82	-15715.61	-15760.07
86	-16323.56	-15711.39	-15730.82	-15780.65
87	-17023.16	-16134.45	-16108.18	-16154.73
88	-18763.93	-17220.72	-17084.88	-17128.00
89	-15787.35	-15266.86	-15296.03	-15348.90
90	-14929.75	-14676.62	-14750.37	-14803.20
91	-16394.24	-15953.09	-16013.23	-16057.63
92	-15740.05	-15279.98	-15326.36	-15371.93
93	-15238.49	-14934.82	-15004.84	-15055.35
94	-16894.92	-16171.32	-16180.79	-16224.25
95	-16504.37	-15839.23	-15854.64	-15898.79
96	-18965.89	-17510.58	-17397.59	-17438.68
97	-16162.51	-15729.23	-15786.11	-15831.82
98	-17500.20	-16662.18	-16659.07	-16700.64
99	-17798.07	-17057.86	-17076.40	-17118.68
100	-15777.42	-15388.06	-15451.17	-15496.13
All	-1653190.96	-1586562.32	-1587921.46	-1592611.31

- (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 = 46.190317]

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	386143129/399998671	0.96536
Genealogies	71870379/1600001329	0.04492

MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
Θ_1	0.60666	2460748.21
Genealogies	0.27805	5665418.34

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