## **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 01:34:38 2017

Program finished at Sun Aug 13 02:14:11 2017 [Runtime:0000:00:39:33]



### **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) 1372961890

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]

Number of concurrent chains (replicates) [c]

Markov chain settings:

Long chain

Number of chains1Recorded steps [a]50000Increment (record every x step [b]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.0.5

Haplotyping is turned on:

Output file: outfile\_0.5\_1.0

Posterior distribution raw histogram file: bayesfile

Raw data from the MCMC run: bayesallfile\_0.5\_1.0

Print data:

Print genealogies [only some for some data type]:

### Data summary

Data file: infile.0.5
Datatype: Sequence data
Number of loci: 100

Mutatio	nmodel:			
Locus S	Sublocus	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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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	
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# Bayesian Analysis: Posterior distribution table

Locus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
1	$\Theta_1$	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00026
2	$\Theta_1$	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00026
3	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
4	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
5	$\Theta_1$	0.00000	0.00007	0.00077	0.00147	0.00287	0.00137	0.00079
6	$\Theta_1$	0.00000	0.00000	0.00010	0.00100	0.00240	0.00103	0.00031
7	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00028
8	$\Theta_1$	0.03420	0.04467	0.04770	0.04913	0.05140	0.04483	0.06991
9	$\Theta_1$	0.00000	0.00000	0.00023	0.00107	0.00240	0.00110	0.00032
10	$\Theta_1$	0.00033	0.00233	0.00363	0.00493	0.00800	0.00397	0.00408
11	$\Theta_1$	0.00000	0.00000	0.00037	0.00113	0.00247	0.00117	0.00041
12	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
13	$\Theta_1$	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00025
14	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
15	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
16	$\Theta_1$	0.00000	0.00000	0.00023	0.00107	0.00240	0.00110	0.00033
17	$\Theta_1$	0.00000	0.00000	0.00057	0.00120	0.00267	0.00123	0.00059
18	$\Theta_1$	0.00000	0.00033	0.00117	0.00187	0.00333	0.00157	0.00115

19	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00022
20	$\Theta_1$	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00025
21	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
22	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
23	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
24	$\Theta_1$	0.00000	0.00000	0.00037	0.00107	0.00247	0.00110	0.00037
25	$\Theta_1$	0.00000	0.00000	0.00023	0.00107	0.00240	0.00110	0.00032
26	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
27	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
28	$\Theta_1$	0.00000	0.00000	0.00070	0.00127	0.00280	0.00130	0.00067
29	$\Theta_1$	0.00513	0.00700	0.01183	0.02067	0.02867	0.01383	0.01507
30	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
31	$\Theta_1$	0.00000	0.00000	0.00057	0.00120	0.00267	0.00123	0.00060
32	$\Theta_1$	0.00000	0.00007	0.00077	0.00147	0.00293	0.00137	0.00080
33	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00030
34	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
35	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
36	$\Theta_1$	0.00000	0.00000	0.00030	0.00107	0.00247	0.00110	0.00036
37	$\Theta_1$	0.00000	0.00080	0.00170	0.00260	0.00413	0.00197	0.00177
38	$\Theta_1$	0.00000	0.00000	0.00037	0.00107	0.00247	0.00110	0.00040
39	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
40	$\Theta_1$	0.00040	0.00240	0.00377	0.00500	0.00820	0.00403	0.00421
41	$\Theta_1$	0.00000	0.00033	0.00117	0.00193	0.00333	0.00157	0.00119

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
42	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00029
43	$\Theta_1$	0.00000	0.00000	0.00030	0.00107	0.00247	0.00110	0.00038
44	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
45	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
46	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
47	$\Theta_1$	0.00000	0.00000	0.00043	0.00113	0.00253	0.00117	0.00045
48	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
49	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
50	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
51	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
52	$\Theta_1$	0.00140	0.00373	0.00530	0.00707	0.01200	0.00597	0.00635
53	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
54	$\Theta_1$	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00025
55	$\Theta_1$	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00026
56	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
57	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
58	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00029
59	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
60	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
61	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023

62	$\Theta_1$	0.00000	0.00000	0.00043	0.00113	0.00253	0.00117	0.00045
63	$\Theta_1$	0.00047	0.00253	0.00390	0.00527	0.00873	0.00430	0.00446
64	$\Theta_1$	0.00000	0.00000	0.00043	0.00113	0.00247	0.00117	0.00041
65	$\Theta_1$	0.00000	0.00000	0.00063	0.00120	0.00273	0.00123	0.00063
66	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
67	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00030
68	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
69	$\Theta_1$	0.00000	0.00000	0.00063	0.00127	0.00273	0.00130	0.00066
70	$\Theta_1$	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00025
71	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00029
72	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
73	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00027
74	$\Theta_1$	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00026
75	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
76	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
77	$\Theta_1$	0.00000	0.00087	0.00183	0.00273	0.00433	0.00210	0.00191
78	$\Theta_1$	0.00000	0.00000	0.00003	0.00100	0.00233	0.00103	0.00030
79	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
80	$\Theta_1$	0.00000	0.00000	0.00030	0.00107	0.00247	0.00110	0.00039
81	$\Theta_1$	0.00000	0.00007	0.00077	0.00140	0.00287	0.00130	0.00077
82	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
83	$\Theta_1$	0.00000	0.00000	0.00070	0.00133	0.00280	0.00130	0.00069
84	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023

_ocus	Parameter	2.5%	25.0%	Mode	75.0%	97.5%	Median	Mean
85	$\Theta_1$	0.00000	0.00000	0.00003	0.00093	0.00227	0.00097	0.00025
86	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
87	$\Theta_1$	0.00000	0.00040	0.00117	0.00193	0.00333	0.00157	0.00119
88	$\Theta_1$	0.00000	0.00000	0.00050	0.00120	0.00260	0.00123	0.00054
89	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
90	$\Theta_1$	0.00000	0.00000	0.00057	0.00120	0.00267	0.00123	0.00059
91	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
92	$\Theta_1$	0.00000	0.00000	0.00037	0.00113	0.00247	0.00117	0.00041
93	$\Theta_1$	0.00000	0.00000	0.00050	0.00113	0.00253	0.00117	0.00048
94	$\Theta_1$	0.00287	0.00560	0.00750	0.00987	0.01700	0.00857	0.00921
95	$\Theta_1$	0.00000	0.00000	0.00070	0.00133	0.00280	0.00130	0.00071
96	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
97	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
98	$\Theta_1$	0.00000	0.00000	0.00030	0.00107	0.00247	0.00110	0.00037
99	$\Theta_1$	0.00000	0.00000	0.00003	0.00087	0.00220	0.00090	0.00023
100	$\Theta_1$	0.00000	0.00073	0.00177	0.00260	0.00420	0.00203	0.00179
All	$\Theta_1$	0.00000	0.00000	0.00023	0.00100	0.00227	0.00103	0.00022

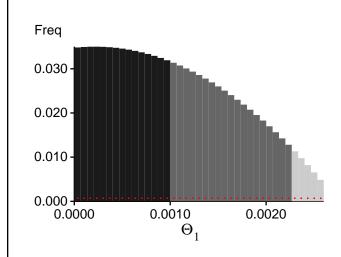
Citation suggestions:

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	-14134.38	-13785.36	-13789.65	-13881.25
2	-14138.32	-13786.54	-13790.20	-13880.93
3	-14143.24	-13792.55	-13798.71	-13888.75
4	-14123.67	-13774.10	-13777.84	-13869.89
5	-14395.46	-14047.91	-14076.10	-14148.25
6	-14153.26	-13806.02	-13812.53	-13901.72
7	-14154.97	-13801.04	-13808.78	-13896.87
8	-34659.85	-26940.76	-25740.91	-25771.10
9	-14189.13	-13828.79	-13839.58	-13923.81
10	-15034.47	-14593.55	-14624.53	-14683.09
11	-14258.18	-13883.10	-13896.00	-13977.44
12	-14124.73	-13774.26	-13777.33	-13869.96
13	-14137.45	-13786.50	-13789.66	-13880.95
14	-14127.79	-13774.00	-13776.84	-13869.26
15	-14119.93	-13771.26	-13774.87	-13867.76
16	-14172.85	-13825.80	-13837.62	-13922.55
17	-14223.23	-13876.25	-13891.76	-13972.82
18	-14350.13	-13991.43	-14018.83	-14088.71
19	-14121.58	-13774.05	-13777.69	-13871.17
20	-14137.42	-13786.82	-13789.74	-13883.16
21	-14126.05	-13774.52	-13777.96	-13872.38
22	-14121.70	-13773.15	-13776.36	-13869.91
23	-14132.46	-13784.58	-13790.98	-13881.55
24	-14213.20	-13851.26	-13862.07	-13944.32
25	-14185.45	-13831.32	-13842.50	-13926.85
26	-14125.18	-13772.91	-13776.17	-13868.51
27	-14122.31	-13773.08	-13776.22	-13869.59
28	-14276.34	-13920.90	-13942.81	-14019.07
29	-17507.18	-16597.55	-16571.09	-16617.74

Migrate 5.0.0a: (http://popgen.sc.fsu.edu) [program run on 01:34:38]

				A010 23
30	-14123.25	-13774.14	-13777.93	-13871.88
31	-14297.47	-13928.29	-13947.42	-14023.04
32	-14436.38	-14030.63	-14046.48	-14118.74
33	-14141.86	-13794.78	-13801.16	-13893.29
34	-14124.72	-13774.25	-13776.96	-13870.31
35	-14127.35	-13774.90	-13777.92	-13871.10
36	-14166.09	-13815.49	-13826.49	-13911.69
37	-15116.01	-14553.29	-14550.24	-14615.67
38	-14168.79	-13819.13	-13828.08	-13915.20
39	-14125.81	-13773.05	-13775.67	-13869.23
40	-15624.89	-14995.29	-14992.04	-15051.92
41	-14735.94	-14250.52	-14258.10	-14325.61
42	-14174.85	-13815.51	-13822.56	-13911.15
43	-14173.06	-13821.14	-13830.43	-13915.92
44	-14146.93	-13794.53	-13801.47	-13892.38
45	-14118.42	-13771.96	-13775.73	-13872.07
46	-14123.45	-13773.78	-13776.79	-13870.42
47	-14180.96	-13829.63	-13842.01	-13925.34
48	-14119.20	-13772.68	-13776.76	-13869.87
49	-14119.17	-13771.34	-13775.38	-13867.94
50	-14148.34	-13793.74	-13799.54	-13888.90
51	-14122.48	-13771.85	-13775.30	-13867.60
52	-15326.63	-14873.12	-14906.34	-14961.74
53	-14123.42	-13774.07	-13777.23	-13870.63
54	-14136.26	-13787.02	-13789.89	-13881.93
55	-14133.99	-13786.33	-13790.71	-13882.42
56	-14123.06	-13773.99	-13777.69	-13869.88
57	-14117.76	-13770.44	-13774.14	-13867.53
58	-14146.80	-13797.36	-13803.03	-13893.82
59	-14134.75	-13786.52	-13792.80	-13884.92
60	-14124.18	-13773.38	-13776.99	-13869.00
61	-14122.13	-13772.88	-13775.30	-13868.76
62	-14203.42	-13848.09	-13861.56	-13943.16
63	-14831.66	-14455.21	-14495.64	-14555.43
64	-14246.55	-13878.03	-13892.74	-13972.85
65	-14258.45	-13899.06	-13919.13	-13995.63
66	-14122.73	-13773.08	-13776.39	-13870.39
67	-14149.80	-13798.61	-13803.93	-13895.58
68	-14121.20	-13771.24	-13774.44	-13867.06
69	-14253.01	-13895.43	-13915.05	-13991.47
70	-14136.02	-13787.59	-13790.94	-13882.20
71	-14166.13	-13813.94	-13822.18	-13909.21
72	-14121.26	-13773.08	-13775.64	-13869.86
73	-14134.41	-13786.63	-13792.58	-13883.48
74	-14134.06	-13786.45	-13790.71	-13882.20

75	-14122.16	-13772.40	-13775.76	-13868.72
76	-14122.32	-13772.46	-13775.86	-13868.15
77	-15066.31	-14583.76	-14599.75	-14666.08
78	-14153.28	-13804.26	-13811.39	-13901.10
79	-14121.28	-13771.84	-13775.81	-13868.08
80	-14154.20	-13809.11	-13821.27	-13906.87
81	-14320.20	-13956.17	-13979.54	-14052.29
82	-14122.52	-13771.08	-13774.15	-13871.05
83	-14205.15	-13860.68	-13881.11	-13958.89
84	-14121.54	-13771.00	-13774.17	-13867.14
85	-14137.10	-13786.05	-13789.29	-13883.34
86	-14122.88	-13773.32	-13777.05	-13869.26
87	-14802.32	-14288.46	-14290.70	-14359.36
88	-14243.97	-13881.79	-13900.26	-13976.56
89	-14122.19	-13772.44	-13775.41	-13871.92
90	-14223.03	-13869.46	-13888.84	-13967.89
91	-14122.23	-13772.66	-13776.31	-13868.98
92	-14166.86	-13819.14	-13829.55	-13916.05
93	-14237.82	-13883.57	-13902.43	-13981.50
94	-19614.44	-17357.68	-17076.24	-17129.30
95	-14244.24	-13890.29	-13910.21	-13986.24
96	-14122.07	-13772.26	-13776.00	-13868.75
97	-14123.33	-13773.91	-13777.92	-13870.05
98	-14175.97	-13829.37	-13839.99	-13925.76
99	-14120.88	-13772.77	-13776.61	-13871.20
100	-14574.71	-14168.80	-14193.59	-14258.45
All	-1452910.81	-1406680.34	-1406018.85	-1414495.56

- (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 = 445.199287]

#### 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	49108695/399993095 821350875/1600006905	0.12277 0.51334

## MCMC-Autocorrelation and Effective MCMC Sample Size

Parameter	Autocorrelation	Effective Sampe Size
$\Theta_1$ Genealogies	0.07758 0.10754	8598661.68 8159091.27

## 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